Page 1

## NATIONAL QUALITY FORUM

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ALL CAUSE ADMISSIONS AND READMISSIONS STEERING COMMITTEE

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TUESDAY May 6, 2014

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The Steering Committee met at the National Quality Forum, 9th Floor Conference Room, 1030 15th Street, N.W., Washington, D.C., at 8:00 a.m., Bruce Hall and Sherrie Kaplan, Co-Chairs, presiding.

PRESENT:

Hospital

BRUCE HALL, MD, PhD, MBA, Co-Chair SHERRIE KAPLAN, PhD, Co-Chair KATHERINE AUGER, MD, Msc, Cincinnati Children's Hospital FRANK BRIGGS, PharmD, MPH, West Virginia University Healthcare JO ANN BROOKS, PhD, RN, Indiana University System JOHN BULGER, DO, MBA, Geisinger Health System MAE CENTENO, DNP, RN, CCRN, CCNS, ACNS-BC, Baylor Health Care System HELEN CHEN, MD, Hebrew Senior Life ROSS EDMUNDSON, MD, Adventist Health System W. WESLEY FIELDS, MD, FACEP, CEP America STEVEN FISHBANE, MD, North Shore University Hospital and LIJ Medical Center LAURENT GLANCE, MD, University of Rochester ANTHONY GRIGONIS, PhD, Select Medical LESLIE KELLY HALL, Healthwise PAUL HEIDENREICH, MD, MS, FACC, FAHA, Stanford University School of Medicine KAREN JOYNT, MD, MPH, Brigham and Women's

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Page 2
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PAULA MINTON-FOLTZ, RN, MSN, Harborview Medical Center; UW Medicine CAROL RAPHAEL, MPA, Subject Matter Expert PAMELA ROBERTS, PhD, MSHA, ORT/L, SCFES, CPHQ, Cedars-Sinai Medical Center FAOTA, ALISON SHIPPY, MPH, Consumer-Purchaser Alliance, National Partnership for Women & Families THOMAS SMITH, MD, FAPA, American Psychiatric Association RONALD STETTLER, United Health Group CRISTIE TRAVIS, MHA, Memphis Business Group on Health NQF STAFF: CHRISTINE CASSEL, MD, NQF CEO TAROON AMIN, Special Assistant to the President and CEO HELEN BURSTIN, Senior Vice President, Performance Measurement ANNE HAMMERSMITH, General Counsel ANDREW LYZENGA, Senior Project Manager, Performance Measurement ADEELA KHAN, Project Manager, Performance Measurement KAREN PACE, PhD, RN, Senior Director, Performance Measurement ZEHRA SHAHAB, Project Analyst ALSO PRESENT: SUSANNAH BERNHEIM, MD, MHS, Yale University JEPTHA CURTIS, MD, Yale University NIHAR DESAI, MD, MPH, Yale University ELIZABETH DRYE, MD, MS, Yale University JANE HAN, Society of Thoracic Surgeons\* LEIN HAN, PhD, CMS LORI GEARY, MPH, Yale University\* JEFF JACOBS, MD, FACS, FACC, FCCP, All Children's Hospital MARI NAKAMURA, MD, MPH, Boston's Children Hospital

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Page 3
JOHN MULDOON, MHA, 3M Health Information
      Systems*
SEAN O'BRIEN, PhD, Duke University
ROBERT MCNAMARA, MD, MHS, Yale University
ISURU RANASINGHE, MD, PhD, Yale University
DAVE SHAHIAN,, MD, Society of Thoracic
Surgeons*
MARK SCHUSTER, MD, MPH, Center of Excellence
      Pediatric Quality Measurement*
for
LARA SLATTERY, MHS, American College of
      Cardiology
LISA SUTER, MD, Yale Center for Outcomes,
      Research and Evaluation
ALAN ZASLAVSKY, PhD, Harvard Medical School
LEORA HORWITZ, MD, MHS, Yale University
* present by teleconference
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Page 4
                   A-G-E-N-D-A
Welcome, Recap of Day 1 . .
                                            6
Consideration of Candidate Measures
(continued)
2514: Risk-adjusted Coronary Artery Bypass
Graft (CABG) Readmission Rate (Society of
Thoracic Surgeons)
                                               18
2515: Hospital 30-day All-cause Unplanned
Risk-standardized Readmission Rate (RSRR)
Following Coronary Artery Bypass Graft (CABG)
                                               57
Surgery (Yale)
2513:
       Hospital 30-day All-cause Risk-
standardized Readmission Rate (RSRR) Following
Vascular Procedure (Yale)
                                               99
       Pediatric All-Condition Readmission
2393:
Measure (Center of Excellence for Pediatric
Quality Measurement)
                                              131
2414:
       Pediatric Lower Respiratory Infection
Readmission Measure (Center of Excellence for
                                              191
Pediatric Quality Measurement)
                                              211
NQF Member and Public Comment
      Hospital 30-day Risk-standardized
0695:
Readmission Rates Following Percutaneous
Coronary Intervention (PCI)(American College
of Cardiology)
                                              222
0505:
       Hospital 30-day All-cause Risk-
standardized Readmission Rate (RSRR) Following
Acute Myocardial Infarction (AMI)
                                              265
Hospitalization (Yale)
       Facility 7-day Risk-standardized
2539:
Hospital Visit Rate after Outpatient
                                              288
Colonoscopy (Yale)
```

	Page 5
Review of Dry Run Results for NQF Measure 1789: Hospital-Wide, All-cause Unplanned Readmissions	326
Next Steps / Committee Timeline	344
NQF Member and Public Comment	349
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	Page 6
1	P-R-O-C-E-E-D-I-N-G-S
2	7:57 a.m.
3	MR. AMIN: Good morning, everyone.
4	Welcome back to day 2 of the Readmissions In-
5	Person Steering Committee meeting.
6	Thank you again for all of your
7	work yesterday. It was quite a day reviewing
8	all the measures that we got through but we
9	were successful.
10	I wanted to welcome Christine
11	Cassel, our CEO at NQF for a quick welcome to
12	the committee.
13	DR. CASSEL: Thanks, Taroon. And
14	thanks, Sherrie and to Bruce too for chairing
15	this important work.
16	I was able to sit in and eavesdrop
17	on some of your conversation yesterday. And
18	so really especially appreciate the
19	thoughtfulness and care and openness, the
20	spirit really of open debate and discussion
21	that goes on here at NQF.
22	And it was what makes the

	Page 7
1	deliberations but also the conclusions of this
2	work carry the weight that they do. Not only
3	in government programs but increasingly in
4	private sector programs as well.
5	So, we just always need to stop
6	just for a moment and recognize all of the
7	volunteers and the expertise around this table
8	that has made the commitment to actually make
9	this multi-stakeholder model of healthcare
10	quality measurement actually work for the
11	nation. So we really appreciate that.
12	I wanted to just I take every
13	opportunity I can to meet with our committees
14	and talk with you about your work but also to
15	let you know what's in a nutshell happening at
16	NQF.
17	We are as you probably have seen
18	recently right in the center of lots of really
19	important discussions and debates about what
20	our board is calling measurement science,
21	really at the forefront of measurement
22	science. Just in the last few weeks all the

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	Page 8
1	attention to whether or not and if so how to
2	risk-adjust measurement for sociodemographic
3	status I think has ignited a needed national
4	debate.
5	Carol Raphael and I were talking
6	about this on the elevator, that these issues,
7	many of these issues just are never done.
8	We're just marching towards a better state, a
9	better measurement science and a better way of
10	implementing quality measurement. And I think
11	that's a really good example.
12	We also are internally at NQF
13	moving to a way of recognizing that endorsed
14	measures there's more to say about whether
15	a measure should be endorsed or not.
16	Sometimes it's not just, you know, an up or
17	down decision. But it's a kind of it depends.
18	And often it depends on what it's being used
19	for and whether it's for trial use.
20	We haven't really had the ability
21	to make those kind of distinctions because we
22	haven't really had the ability to do rapid

Page 9 1 cycle feedback and follow-up. And we now are actually moving to 2 a place where we can do that with an open 3 pipeline for measure submission, a model of 4 standing committees so that it doesn't take us 5 six months to assemble a committee every time 6 we have a group of new measures. 7 8 We're not there yet, but with the 9 full support of CMS we are piloting some of 10 these approaches with some of the contracts we 11 have this year. And I'm very hopeful and optimistic that it's going to give us a way of 12 13 reducing the cost and the length of time that it takes to get this process through and to be 14 able to be more adaptive then to what's 15 happening in the scientific world as well as 16 17 to what the clinical world really needs from 18 us. Another thing that is happening is 19 20 that you know we have a process to evaluate the measures and endorse them or not. 21 And 22 then we have a process called MAP, the Measure

	Page 10
1	Applications Partnership, which selects which
2	measures should be used for which federal
3	programs.
4	And a lot of the discussion that
5	goes on at the MAP is very similar to this
6	discussion that goes on here. And in many
7	ways we think it could be much more efficient
8	if the two were sort of streamlined more to be
9	able to identify what purpose maybe at this
10	stage what purpose measures should be used for
11	and feed that into the MAP process so that it
12	makes that whole process more streamlined.
13	And in order to do that within the
14	staff we're taking advantage of staff
15	expertise in Lean and Six Sigma reengineering.
16	I don't know if you realize that
17	Taroon is a black belt among us here. So, in
18	his spare time he's actually helping us with
19	the staff process of finding ways to cut the
20	waste out of our process and become more lean
21	and more efficient and effective in what we
22	do. So we're very fortunate to have that kind

	Page 11
1	of expertise on our staff as well.
2	So you'll be hearing more about
3	many of these things that are developing at
4	NQF. But in the meantime the policies are the
5	policies and it's really important that we
6	adhere to these publicly available and
7	carefully established policies which you're
8	helping us to implement here today.
9	So let me just stop there,
10	Sherrie, and see if anybody has any quick
11	questions. I know you have a lot of work to
12	do so I don't want to hold you up too long.
13	CO-CHAIR KAPLAN: Questions or
14	comments for Chris? Dr. Cassel.
15	DR. CASSEL: Okay, thank you.
16	CO-CHAIR KAPLAN: Well, thank you.
17	That was very, very helpful, especially the
18	knowing that on the horizon there may be some
19	of the synergy between the MAP process and
20	this process I think helps with some of the
21	frustrations that many feel for the purpose of
22	measurement and a little concern about how

	Page 12
1	these things are going to be applied. So that
2	was very, very helpful. Thanks, Chris.
3	So, I'm going to ask Taroon to
4	kick off in a second, but yesterday it was
5	noted that we laid out a bunch of ground rules
6	and then assiduously ignored all of them.
7	And so today we are going to
8	adhere to some ground rules hopefully more
9	closely. We did a lot of hard work and you
10	were extremely efficient but I think it will
11	help if we can make the questions and the
12	comments very concise. And then hopefully
13	from the measure developers also the responses
14	concise. And so that will keep us on track to
15	make sure because we're going to start
16	losing people as their plane flights and so on
17	get going.
18	So if we can crisp up our comments
19	and questions, and then also the responses
20	from the measures developers, if those can be
21	a little bit more concise will get us through
22	today's agenda in a timely and hopefully full-

	Page 13
1	throated discussion way.
2	So, Taroon?
3	MR. AMIN: So, if we can as we
4	get started here I would welcome the
5	developers from STS to join us at the table as
6	we get started.
7	But I'll actually turn it over to
8	Adeela if you can just walk through a quick
9	summary of day 1 here and then just quickly
10	walk through the agenda for day 2.
11	MS. KHAN: Sure. So we actually
12	were able to get through the agenda for day 1
13	and evaluate all the measures that were
14	supposed to be.
15	Just a quick recap of where each
16	measure is. 2502, the all-cause unplanned
17	readmission measure for 30 days post discharge
18	from inpatient rehab facilities passed.
19	2512, all-cause unplanned
20	readmission measure for 30 days post discharge
21	from long-term care hospitals was a measure
22	where we weren't able to reach consensus. And

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	Page 14
1	so we'll be following up with those developers
2	but that measure will still be going out for
3	public and member comment.
4	2375, PointRight OnPoint 30 SNF
5	rehospitalizations passed.
6	2510, skilled nursing facility 30-
7	day all-cause readmission measure passed.
8	2496, standardized readmission
9	ratio for dialysis facilities was another
10	measure where consensus was not reached.
11	2503, hospitalizations per 1,000
12	Medicare fee-for-service beneficiaries was
13	consensus not reached.
14	2504, 30-day rehospitalizations
15	per 1,000 Medicare fee-for-service
16	beneficiaries was consensus not reached.
17	2505, emergency department use
18	without hospital readmission during the first
19	30 days of home health passed.
20	2380, rehospitalization during the
21	first 30 days of home health passed.
22	And 0327, risk-adjusted average

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	Page 15
1	length of inpatient hospital stay was another
2	measure where consensus was not reached.
3	Quickly we're going to be going
4	over the following hospital measures today.
5	We have two CABG measures, one risk-adjusted
6	vascular procedure, two pediatric measures,
7	one all-cause, one lower respiratory
8	infection. And we have a PCI, AMI and
9	outpatient colonoscopy.
10	MR. AMIN: Okay, thanks. Thanks,
11	Adeela. I just wanted to point out two other
12	things for today.
13	Again, a sincere try all for
14	yesterday. There was a lot of work that was
15	done. Again, NQF would not be able to achieve
16	its goals without volunteers like you spending
17	time through the workgroup calls and time that
18	we had yesterday and today.
19	I just wanted to also follow up
20	that we'll try to do an evaluation or just a
21	discussion around the dry run results on the
22	1789 all-cause measure today.

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	Page 16
1	And given the recommendation for
2	endorsement on the two measures yesterday that
3	were had the same measure focus and the
4	same target population we will likely have a
5	follow-up call to have a discussion around
6	best-in-class/competing in addition to the
7	potential two measures that are up for
8	discussion today.
9	But we'll first evaluate them
10	independently and then have that discussion
11	later on.
12	So again, I'll turn it over to the
13	chairs if there's any other reflections on
14	yesterday. Otherwise, you guys can lead us
15	directly into the conversation.
16	Is there anything else you wanted
17	to add, Helen? Okay, thank you.
18	CO-CHAIR KAPLAN: And my earlier
19	comments were not meant to truncate anything.
20	I just want to make sure everybody gets a
21	chance to say what they have to say about the
22	measures under consideration and do so and get

Page 17 1 their points made and considered. With respect to the reviewers, if 2 3 you could each time you make a comment state your name because the people recording this 4 can't see you. And they know which 5 microphones we're sitting at so they can hear 6 who we are but they can't hear who you are. 7 So if you could each time say your 8 9 We're going to ask yourself to name. 10 introduce yourself and your background quickly 11 and then give a two-minute summary of the 12 measure. 13 But if you could, when you're talking, after your introductions repeat your 14 name again that will be very helpful. 15 So 16 could you introduce yourselves and then the 17 measure. My name is Jeff 18 DR. JACOBS: Jacobs and I'm a cardiac surgeon from Johns 19 20 Hopkins and All Children's Hospital in St. 21 Petersburg, Florida. And I'm joined today by Sean O'Brien. 22

	Page 18
1	DR. O'BRIEN: Hi, my name's Sean
2	O'Brien. I'm a statistician at Duke
3	University Medical Center. And we serve as an
4	analytic center, the STS database.
5	DR. JACOBS: We also have on the
6	phone with us two members from the Society of
7	Thoracic Surgeons, Dave Shahian who's a
8	cardiac surgeon from Harvard, and Jane Han who
9	is staff at Society of Thoracic Surgeons.
10	CO-CHAIR KAPLAN: Cathy, can we
11	just make sure that those people have open
12	lines?
13	OPERATOR: Yes, ma'am, their lines
14	are open.
15	DR. JACOBS: So, the measures that
16	are before the group today, actually the next
17	two were developed in parallel. There's the
18	Risk-adjusted Coronary Artery Bypass Graft
19	Readmission Rate from the Society of Thoracic
20	Surgeons and there's the Hospital 30-day All-
21	cause Unplanned Risk-standardized Readmission
22	Following Coronary Artery Bypass Graft Surgery

	Page 19
1	from Yale.
2	These two measures were developed
3	in parallel and in collaboration between the
4	two groups as part of a process in
5	collaboration with CMS.
6	And the two measures are somewhat
7	different in that the Yale measure is based on
8	administrative data and the Society of
9	Thoracic Surgeons measure is based on clinical
10	data from a clinical database, a clinical
11	database that currently has a penetrance of
12	between 90 and 95 percent of hospitals in the
13	United States, but not 100 percent.
14	And I think both groups, the Yale
15	group and STS, view these two measures as
16	complementary with strengths and weaknesses
17	that complement the other measures.
18	And I think the way I kind of look
19	at this simplistically is that the Yale
20	measure is based on administrative data. An
21	advantage and a strength of the Yale measure
22	is that 100 percent of hospitals in the United

	Page 20
1	States that do coronary artery bypass surgery
2	participate in the administrative data sets
3	which the Yale measure is developed.
4	Meanwhile, in the Society of
5	Thoracic Surgeons only 90-95 percent of the
6	hospitals participate. But we feel that the
7	actual risk adjustment in the measure is
8	somewhat enhanced because of increased ability
9	to use clinical variables both for defining
10	the model of patients, the cohort of patients,
11	isolated CABG patients, and for the variables
12	used to risk-adjust.
13	I think that's a brief summary of
14	how these two measures fit together and what
15	we've done. And I think we'd be happy to
16	answer any questions.
17	CO-CHAIR KAPLAN: Thank you very
18	much. Bruce and Paul were the discussants on
19	this measure. With respect to the evidence,
20	Bruce, do you want to go first?
21	CO-CHAIR HALL: I think overall
22	this is an outstanding measure and in terms of

	Page 21
1	evidence the evidence is strong.
2	DR. HEIDENREICH: I would agree.
3	While I think it should be classified as an
4	intermediate clinical outcome I think there's
5	a strong rationale.
6	CO-CHAIR KAPLAN: Other comments
7	and questions from the group? Are we ready to
8	vote evidence?
9	MS. SHAHAB: So, we're going to
10	vote on 1(a) evidence, 1 for yes, 2 for no.
11	And your time begins now. I think we have all
12	22 votes. 1(a) evidence, 22 yes, zero no.
13	CO-CHAIR KAPLAN: Thank you very
14	much. Onto performance gap.
15	CO-CHAIR HALL: I'll jump in
16	front, Paul, and then you can follow.
17	Overall, about a 13.5 percent readmission rate
18	depending on exactly which set and period and
19	so on that you look at.
20	In terms of risk-standardized rate
21	it comes out to about 17 percent with a range
22	of 12.5 to 34.2. The interquartile being

Page 22 1 about 15 to 18. So, about a 3 percentage point 2 3 spread in the interquartile range. So, at least, you know, comparable if not as good or 4 better than many of the other measures that 5 we're dealing with. 6 In terms of the 7 performance gap. 8 DR. HEIDENREICH: I would agree. 9 I would say it's at least a moderate level of 10 potential for improvement. 11 CO-CHAIR KAPLAN: Other comments from the committee? Hearing none are we ready 12 13 to vote? MS. SHAHAB: Voting for 1(b) 14 performance gap, 1 high, 2 moderate, 3 low, 4 15 insufficient. Time begins now. 16 Just one 17 more. We have all 22 votes for 1(b) 18 performance gap: 6 high, 16 moderate, zero 19 20 low, zero insufficient. CO-CHAIR KAPLAN: Great. 21 For scientific acceptability we're going to move 22

	Page 23
1	to reliability first. Sorry, priority.
2	Priority.
3	CO-CHAIR HALL: In terms of
4	priority the developers make the case that
5	it's between two and three hundred million
6	dollar target roughly as a subset of all
7	readmission costs that are a burden on
8	Medicare.
9	So, again, I think it's that's
10	probably an understatement of the magnitude of
11	the issue in terms of priority. So, my
12	feeling on priority was at least moderate if
13	not better.
14	CO-CHAIR KAPLAN: Paul?
15	DR. HEIDENREICH: I would agree on
16	moderate. It's not as common as readmission
17	as some of the other medical diagnoses, but
18	still substantial.
19	CO-CHAIR KAPLAN: Other comments?
20	Are we ready to vote on priority?
21	MS. SHAHAB: Voting for 1(c) high
22	priority, 1 high, 2 moderate, 3 low, 4

	Page 24
1	insufficient and your time begins now.
2	We have all 22 votes, 1(c) high
3	priority. Five voted high, seventeen
4	moderate, zero low, zero insufficient.
5	CO-CHAIR KAPLAN: Thank you. Now
6	onto scientific acceptability and first
7	reliability.
8	CO-CHAIR HALL: So I think the
9	developers have provided outstanding
10	information. I think in terms of reliability
11	actual implementation cutoffs for distinction
12	and whatnot are actually not described. Those
13	would be determined at a later date. So in
14	terms of the most kind of classic rigorous
15	signal-to-noise it's not really possible to
16	comment on that until some of those later
17	details would be specified.
18	In the background I think every
19	other aspect of data field reliability,
20	consistency, reproducibility, and so on are
21	met by what has been submitted in the
22	material.

	Page 25
1	CO-CHAIR KAPLAN: Paul?
2	DR. HEIDENREICH: I did see, maybe
3	you can correct me, but it looks like you did
4	compare, say, one year and three years worth
5	of data. If I got that right. And that had
6	reasonable correlations. It looks like it
7	looked like you had a score-out of 0.78 if you
8	had 300 hospitals included. I don't know if
9	you have any further clarification.
10	DR. O'BRIEN: This is Sean O'Brien
11	from Duke University.
12	One thing we did with respect to
13	reliability was to estimate the proportion of
14	variation that was explained by true signal
15	variation as opposed to random statistical
16	fluctuations.
17	And for that type of analysis we
18	basically used the sample of all of the
19	hospitals, there's approximately 1,000
20	hospitals in the development data set.
21	And when you included all the
22	hospitals that have at least 30 cases which

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	Page 26
1	might be a very common, inclusive threshold
2	for reporting results of a hospital, the
3	reliability, the percentage of explained
4	variation by signal was 47 percent.
5	And then we looked at thresholds
6	of what if we only reported results for
7	subsets of hospitals that have at least 50
8	cases, at least 100, or at least 200 and those
9	are respectively around 50 percent, 55 percent
10	and 65 percent.
11	So I think the results
12	demonstrated the potential for higher
13	reliability with even with a very inclusive
14	threshold as being kind of, you know, adequate
15	or moderate, and the potential to have very
16	high reliability in the subset with larger
17	volumes.
18	In terms of the comparing one year
19	versus three year, that may be I think
20	basically we saw high agreement between
21	different outcomes done across different time
22	periods. But actually I don't have the

	Page 27
1	details up in front of me.
2	CO-CHAIR HALL: So I'd like to
3	support but back up and clarify something that
4	Sean said.
5	So indeed, Paul, as you point out
6	there was good consistency and reproducibility
7	of fields and data.
8	With respect to the reliability
9	that you talk about, Sean, again, I respect
10	your numbers. When an institution submits a
11	30-case estimate you had a number of 0.47.
12	For 200 cases it would be 0.641. The
13	development was on a 3-year data set, right?
14	So, again, if that's how the
15	measure were going to be implemented then
16	those would be good reflections of the
17	reliability.
18	But those the actual sort of
19	application of the measure is not yet
20	determined. And so, again, those are good
21	numbers. As Sean correctly phrased it that
22	represents great potential for high

	Page 28
1	reliability. And so I support what he said.
2	But those numbers reflect a 3-year
3	data set and an example of a development, an
4	example of the potential that could be
5	achieved.
6	CO-CHAIR KAPLAN: Other comments
7	from the group? Go ahead, Larry.
8	DR. GLANCE: So, I have a comment
9	about specification. I have great respect for
10	the fact that there is a tremendous amount of
11	clinical expertise as well as statistical
12	expertise that went into the development of
13	this outstanding measure.
14	The comment that I have is that
15	this is a measure for isolated CABG surgery.
16	And as part of the specification it also
17	includes patients who underwent a combined
18	CABG and ventricular assist device placement.
19	The rationale as I understand it
20	for including the VAD patients is that there
21	are occasions where because of a quality issue
22	a patient is unable to separate from the

	Page 29
1	heart-lung machine and requires a ventricular
2	assist device.
3	On the other hand at many heart
4	failure centers there is an intention going
5	into the surgical procedure that these are
6	very, very high-risk patients and that there
7	is a high likelihood that the patient will in
8	fact, although the planned procedure is a
9	CABG, that the patient will in fact probably
10	need to undergo ventricular assist device.
11	The reason this is important is
12	because the readmission rate for CABG patients
13	is very, very different from the readmission
14	rates for ventricular assist device patients.
15	So my question for the developers is, knowing
16	this, why would you have included VAD patients
17	as part of the specification for this measure.
18	DR. JACOBS: Well, thank you.
19	This is Jeff Jacobs. And first of all, that's
20	an excellent question and some excellent
21	observations.
22	A couple of clinical facts that I

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	Page 30
1	think would help understand the rationale for
2	the way this was developed.
3	First of all, substantially less
4	than 1 percent of all coronary artery bypass
5	grafts performed in the United States are
6	associated with the use of a ventricular
7	assist device. Ninety-nine percent of them
8	are not.
9	Of those that are associated with
10	ventricular assist device usage most of them
11	are unplanned. And a patient is taken to the
12	operating theater, undergoes coronary artery
13	bypass grafting, cannot separate from the
14	bypass machine, a variety of interventions are
15	tried including a machine called an intra-
16	aortic balloon pump.
17	After all of those things failed
18	then really the only option is to put the
19	patient on a ventricular assist device. And
20	most of the time that's in an unplanned
21	situation.
22	It is true that in some heart

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	Page 31
1	failure centers patients go to the operating
2	theater for planned ventricular assist device
3	insertion after coronary artery bypass
4	grafting but that's rare. It's extremely
5	rare.
6	That being said, it is a fact that
7	patients who get a ventricular assist device
8	either planned or unplanned after a coronary
9	artery bypass grafting, should they survive
10	and go home have a higher rate of readmission
11	than those without a ventricular assist
12	device, no doubt.
13	When the measure was developed the
14	measure was developed using a data set where
15	we just knew that the CABG was associated with
16	ventricular assist device insertion.
17	Since that time the STS database
18	has been modified so that ventricular assist
19	devices are now tracked as to whether or not
20	the insertion is planned or unplanned. And
21	that's a change in the database since the
22	measure was developed.

	Page 32
1	Therefore, moving forward we
2	certainly can implement this measure with the
3	definition of isolated CABG that would include
4	patients who had an unplanned ventricular
5	assist device but excluded those with a
6	planned ventricular assist device.
7	CO-CHAIR KAPLAN: Thank you. I
8	think this discussion shades into validity
9	because we're talking then about the accuracy
10	of the application of the measure rather than
11	the reproducibility of it. So, anymore
12	questions on reproducibility? Go ahead, Paul.
13	DR. HEIDENREICH: Well, I have a
14	question about that, but we can hold it.
15	CO-CHAIR KAPLAN: Can we hold it?
16	Yes, for the validity question. Because right
17	now I'd like to stick to reproducibility.
18	So, we've heard the
19	reproducibility is in the zone. And actually
20	for some of us who look at these kinds of
21	signal-to-noise numbers for those hospitals
22	that have a fairly large number that's a good

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	Page 33
1	number. It's a reasonable number to have
2	given the database.
3	Any other comments on
4	reproducibility? Are we ready to vote?
5	MS. SHAHAB: Voting for 2(a)
6	reliability - 1 is high, 2 moderate, 3 low, 4
7	insufficient and your time begins now.
8	We have all the votes for 2(a)
9	reliability. Eight high, fourteen moderate,
10	zero low, zero insufficient.
11	CO-CHAIR KAPLAN: Thank you. Now
12	we're onto validity. So, Bruce, do you want
13	to?
14	CO-CHAIR HALL: I have a couple of
15	points, but Paul, you were about to go ahead.
16	DR. HEIDENREICH: Well, just a
17	follow-up to your comment that you can now
18	have a field for a planned ventricular assist
19	device is how can you are you confident
20	that that won't be gamed? That seems very
21	hard to control.
22	DR. JACOBS: Absolutely. I

Page 34 1 anticipated that being the next question. I've got the word "gaming" written right here. 2 3 (Laughter) DR. JACOBS: Oh, this is Jeff 4 Jacobs. I'm supposed to identify myself. 5 Well, that's a great question. 6 And I think one potential tradeoff of 7 including patients with unplanned ventricular 8 9 assist device and excluding patients with 10 planned ventricular assist devices is that the 11 system then is subject to gaming. I think the way that can be 12 13 addressed is through a combination of proper definitions, proper documentation and then 14 audit of that documentation. 15 The Society of Thoracic Surgeons 16 17 database is one of the most rigorously audited clinical databases in the United States with 18 multiple sites undergoing site visits with 19 20 audit every single year. And it would be a relatively 21 simple process during that audit to audit this 22

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	Page 35
1	field and to make sure that the documentation
2	is in place to document that the patient truly
3	went to the operating theater with a planned
4	ventricular assist device insertion.
5	And that would simply require a
6	note in the chart that says that the family
7	was consented for a planned ventricular assist
8	device and that the clinical team felt it was
9	likely that that might be needed because of
10	the patient's severe heart failure.
11	So I agree with you that gaming is
12	a potential problem. The solutions to
13	addressing that are good definitions,
14	documentation and good audit.
15	CO-CHAIR KAPLAN: Thank you. I
16	think that may get to so Paul, you might
17	want to bring that up again, or at least
18	remind people of it when it comes to the use.
19	Because unintended consequences is one of the
20	use parameters. Bruce?
21	CO-CHAIR HALL: So in terms of
22	reliability I thought I would present my 20-

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	Page 36
1	second summary for the group. This is a risk-
2	standardized readmission ratio. Ninety-five
3	percent intervals are provided. Medicare fee-
4	for-service greater than 65 then obviously but
5	matched to the STS data. So matching between
6	the programs is one of the prominent features.
7	This is isolated CABG. We've
8	already heard some comments about that. The
9	definition of isolated CABG is provided and
10	well specified with note to that issue that
11	we've already heard about.
12	Patients have to be discharged
13	alive and then readmitted within 30 days from
14	discharge.
15	The exclusions include under 65,
16	patients that they were not able to match
17	between the data programs, cases that are not
18	deemed standalone by their definition which
19	we've touched on, patients who died in the
20	hospital or were discharged to AMA.
21	Now, patients who died in the
22	hospital, there's some small controversy about
	Page 37
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1	how to handle a death in the hospital with
2	respect to readmissions because obviously that
3	was not a good outcome and yet that patient is
4	not eligible for readmissions.
5	Suffice to say that because that
6	is controversial and there's probably a lack
7	of 100 percent consensus within healthcare
8	about how to do that at least as many people
9	are excluding deaths as taking any other
10	approach. So this is consistent with that.
11	If patients were not fee-for-
12	service, excluded index more than 365, not
13	first admission, all exclusions, all specified
14	well.
15	Race and sociodemographics were
16	not included but the developers provide good
17	information about those variables. Again,
18	this the information we're seeing is a 3-
19	year development set. If it were a 3-year
20	measure that might be questioned but it's not
21	clear that in practice it would be a 3-year
22	measure.

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	Page 38
1	And the readmission is attributed
2	to the first institution in the case where
3	patients are transferred between acute
4	institutions. And only one readmission would
5	be counted.
6	So I think overall excellent
7	specifications and high validity on all the
8	aspects that I mentioned.
9	CO-CHAIR KAPLAN: Do the
10	developers want to respond to that?
11	DR. JACOBS: Is there a particular
12	question?
13	CO-CHAIR KAPLAN: Never mind.
14	Paul?
15	DR. HEIDENREICH: I'd say the only
16	concern for validity I would have, it's not a
17	big concern, is the matching to CMS. I think
18	given right now you're not allowed to match on
19	Social Security number, is that correct? But
20	obviously CMS could do that in the future.
21	So, I assume there's not a 100 percent
22	matching to the readmission to CMS.

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	Page 39
1	DR. JACOBS: Right, so the issues
2	we've had with Social Security number relate
3	to our matching of the STS database to the
4	Social Security Death Master File.
5	And that matching process worked
6	quite well. We've published several papers
7	and all was well until the Social Security
8	Death Master File was modified. And with
9	changes in the Social Security Death Master
10	File a substantial portion of the deaths in
11	the Social Security Death Master File are no
12	longer re-disclosed. So that's not very
13	useful for outcomes research.
14	As far as our matching with CMS, I
15	think that it's I'll let Sean address the
16	overall numbers but my understanding is that
17	the overwhelming majority of patients over the
18	age of 65 in the STS database are matched
19	successfully to the CMS registry.
20	DR. O'BRIEN: This is Sean
21	O'Brien. Yes, Jeff, that's correct. Using an
22	indirect record linkage around 85 percent are

	Page 40
1	Page 40 linked. But then if you look within the
	_
2	subset of sites that are actively
3	participating it depends which direction
4	you're linking from among the subset in
5	Medicare what percent linked to the STS
6	database within sites that are actively
7	participating in the database, high nineties,
8	97 percent, 98 percent. So it's fairly
9	complete.
10	And then going the other direction
11	of course you don't pick up the Medicare
12	Advantage plans, the HMOs. Of course those
13	wouldn't show up in the claims-based measure
14	either.
15	DR. O'BRIEN: Can I just this
16	is Sean O'Brien again. Just to respond to the
17	question about the time frame.
18	All aspects of the measure were
19	really developed with consistency with other
20	CMS readmission measures in mind. So a lot of
21	the other readmission measures for AMI and
22	pneumonia, et cetera, they were originally

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	Page 41
1	developed with a 1-year time frame and they
2	were subsequently in subsequent iterations
3	converted to a 3-year time frame. So I think
4	the measure developers had a 3-year time frame
5	in mind just for consistency with CMS.
6	CO-CHAIR KAPLAN: Thanks for that
7	clarification. Any other comments on
8	validity? Go ahead.
9	MS. SHIPPY: I had a quick
10	question about the specification. So, you had
11	noted in your introduction that you had worked
12	in collaboration with CMS. Can you discuss
13	the choice for 65 and older for the patient
14	denominator and CMS has 18 and older.
15	DR. O'BRIEN: CMS has 18 and
16	older?
17	MS. SHIPPY: I understand that
18	this is probably harmonization but it felt
19	like it was an opportunity to have them
20	discuss it.
21	DR. JACOBS: My understanding is
22	that to be eligible for Medicare one has to be

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	Page 42
1	either over the age of 65 or in renal failure.
2	So, to be in the Medicare database and having
3	undergone a CABG the two ways to get in there
4	is either being younger and being in renal
5	failure, or being over the age of 65. So
6	that's why the age of 65 is part of this
7	measure.
8	I'm not sure why a claims-based
9	measure would be developed for over the age of
10	18 that's based on the Medicare data because
11	I don't think a patient would be eligible for
12	that unless they're on dialysis.
13	CO-CHAIR KAPLAN: Okay, when you
14	put up your cards if you can turn them
15	sideways so I can see them. They disappear
16	when they're this way. Paul?
17	DR. HEIDENREICH: I know the Yale
18	group has occasionally used the California
19	data to test their model that was developed
20	for 65 and older in Medicare but then to have
21	it on a claims base for 18 and above. So I
22	didn't deal with that measure but I know

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	Page 43
1	they've done that with other measures.
2	CO-CHAIR KAPLAN: Thank you.
3	Other comments? Okay, we're ready to vote
4	validity.
5	CO-CHAIR HALL: So I have a
6	question though. Is are we asking that the
7	developers clarify that in the future the VADs
8	would be so specified as planned or unplanned.
9	Do we have that ability to request that? If
10	they agree?
11	DR. JACOBS: We're very
12	comfortable with that. We have the skills to
13	do it and we're comfortable doing that.
14	CO-CHAIR KAPLAN: So how are we
15	approving the measure though? As specified,
16	as is. We can recommend that they make this
17	change but we are approving or not approving
18	the measure as it is currently presented.
19	Ready to vote? Okay.
20	MS. SHAHAB: Voting for 2(b)
21	validity, 1 high, 2 moderate, 3 low, 4
22	insufficient and your time begins now. Just

Page 44 1 one more vote. We have all the votes for 2(b) 2 3 validity. Four voted high, seventeen voted moderate, one low and zero insufficient. 4 CO-CHAIR KAPLAN: Okay, we're onto 5 feasibility. 6 CO-CHAIR HALL: I thought the 7 8 feasibility was very reasonable. The main 9 issues I guess that popped into mind would 10 again be the penetration of the STS program 11 into all CABG procedures across the country which I think is very, very high. 12 13 And then the issues around matching that have already been raised which 14 might create some limitation around perfect 15 matching. But again, the matching seems to be 16 17 done at a very high level. So otherwise, I did not see any 18 major obstacles to feasibility. I defer to 19 20 Paul. 21 DR. HEIDENREICH: I agree with that. 22

	Page 45
1	CO-CHAIR KAPLAN: Any other
2	comments? Go ahead.
3	MS. HALL: I had a question about
4	the proprietary nature and the potential fees
5	and could that cause barriers for use by other
6	organizations or particularly those who might
7	be reporting to the public for public good or
8	consumer organizations who might have an
9	interest. Could you comment on that, please?
10	DR. JACOBS: So, the Society of
11	Thoracic Surgeons is the largest professional
12	organization of cardiac surgeons in the world.
13	And almost all cardiac surgeons in the United
14	States are members.
15	STS is a strong advocate of public
16	reporting, a huge advocate. And currently our
17	outcome data is publicly reported on two
18	platforms, one through Consumers Report. And
19	we partner with Consumers Report because that
20	allowed public reporting from a respected
21	organization at an arm's length from STS.
22	So, Consumers Report publicly

	Page 46
1	reports our outcomes measures with a web
2	platform designed by Consumers Report and
3	using STS data.
4	STS also reports our outcome data
5	on our own website, www.sts.org. That
6	information is available to anyone through the
7	internet for free. So there's methods to
8	access the results of our NQF-endorsed
9	measures through our website for free and also
10	through Consumers Report at an arm's length
11	from us.
12	So I think the issue of public
13	reporting and the proprietary nature of the
14	database becomes essentially a non-issue
15	because there's two ways to get that
16	information from the website of STS or
17	Consumers Report. And we're certainly a big
18	advocate of transparency in public reporting.
19	CO-CHAIR KAPLAN: Thank you.
20	Other questions or comments? Okay, we're
21	ready to vote.
22	MS. SHAHAB: Voting for number 3,

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	Page 47
1	feasibility. One is high, two moderate, three
2	low, four insufficient and your time begins
3	now.
4	We have all the votes for
5	feasibility. Eleven voted high, eleven
6	moderate, zero low and zero insufficient.
7	CO-CHAIR KAPLAN: Thank you. Now
8	we're on the issue of usability and use.
9	CO-CHAIR HALL: I felt the
10	usability was high myself. I did not see any
11	major obstacles.
12	CO-CHAIR KAPLAN: Paul?
13	DR. HEIDENREICH: There's always
14	the potential that with a procedure that's
15	elective you could have surgeons not doing the
16	cases. Although, as you say, you've been
17	reporting data for a long time so I don't
18	think this would have any significant
19	incremental impact on selecting cases based on
20	risk of readmission.
21	And then it just has the
22	limitation I think that all the CMS 3-year

	Page 48
1	based measures have in that it just takes
2	you can't see a rapid change in your program
3	through those data.
4	CO-CHAIR KAPLAN: So the issue of
5	gaming was raised. And I want to make sure
6	we're all having a discussion about unintended
7	consequences. Karen?
8	DR. JOYNT: Just two quick
9	comments. I think one, just to get back to
10	the ability of these models to sort of reduce
11	the information that you can get from low-
12	volume hospitals. And for two reasons, in
13	case you're looking for a safety signal and
14	also for consumers as we talked about
15	understanding what the difference is between
16	an average hospital that's small and a
17	hospital about which we just don't know their
18	performance.
19	Again, this all gets back to
20	usability as opposed to the validity and I'm
21	sorry to be a broken record on this.
22	And I forget what my other comment

	Page 49
1	was.
2	CO-CHAIR KAPLAN: Thank you. Do
3	the developers want to comment on that? I
4	mean, this is excuse me for interrupting.
5	This is going to be I mean,
6	low-volume hospitals are going to have the
7	same plaguing problem we've all talked about
8	and will probably continue to debate for the
9	majority of our remaining careers.
10	So it is one of these perplexing
11	problems. Do you just not evaluate the low-
12	volume hospitals? Do you evaluate them and
13	give them the mean? Do you do all the things
14	that we've talked about that they're going to
15	always have problems associated with the error
16	of estimation.
17	Developer?
18	DR. JACOBS: This is Jeff Jacobs
19	again. And I think this recent dialogue
20	raised three issues. One, risk aversion as an
21	unintended consequence, two, gaming, and
22	three, how to manage low-volume hospitals.

	Page 50
1	And I'll say a couple of sentences about each
2	one and then I can answer more questions.
3	Certainly with any form of outcome
4	reporting risk aversion is a possibility. The
5	solution to that is a good risk adjustment
6	methodology. And I think this measure as well
7	as all of our isolated CABG measures
8	implemented by STS have a very vigorous risk
9	adjustment methodology that's designed to
10	prevent risk aversion.
11	I think it's extremely unlikely
12	that this particular measure would lead to any
13	new risk aversion because isolated CABG is a
14	group of patients that are already subject to
15	multiple other NQF-endorsed measures including
16	a mortality/morbidity/multi-domain composite.
17	So I think that although risk
18	aversion is possible with any measure it's
19	miigated by proper risk adjustment.
20	Regarding gaming of the system
21	which came into context in this discussion
22	with the planned versus unplanned VAD but

Page 51 1 certainly could also become an issue with other components of any risk adjustment 2 methodology I think the solution to gaming is 3 having, again, good definitions for all the 4 fields, having proper documentation of those 5 definitions, and the application of those 6 definitions and having a solid audit program. 7 And as I said before, I think our 8 audit program of the STS database is as good 9 10 as any and better than most clinical 11 registries in the country. Finally, related to the issue of 12 13 low-volume hospitals, this is a challenging problem for almost any measure, especially 14 measures that deal with relatively rare 15 16 procedures and relatively rare operations. 17 And I think within STS we've done a lot to make sure that appropriate confidence 18 intervals are utilized so that the low-volume 19 20 hospitals and their unique situations are respected and accounted for. 21 And I think I'll turn this over to 22

	Page 52
1	Sean to maybe make an additional comment about
2	what we do to deal with low-volume hospitals.
3	It's a topic we discussed on frequent phone
4	conferences.
5	DR. O'BRIEN: This is Sean
6	O'Brien. I don't think there's a magic bullet
7	for dealing with the problem of small sample
8	sizes.
9	As Jeff mentioned we report
10	measures with measures of uncertainty so I
11	think that's about the best you can do is to
12	say what the evidence is and report that
13	there's a range of possible performance that's
14	consistent with the observed data.
15	CO-CHAIR KAPLAN: Thank you.
16	Bruce?
17	CO-CHAIR HALL: So I would like to
18	add that with respect to this particular
19	measurement access isolated CABG the
20	developers do shed light on this for us in
21	their reliability information.
22	For instance, portraying that the

	Page 53
1	signal-to-noise version of the reliability
2	assessment down to 30 cases gives a metric
3	about 0.47.
4	So with 30 cases assessed that
5	level of reliability is as good or better than
6	probably anything you see in healthcare which
7	I don't know if that's a reflection that
8	isolated CABGs end up being a pretty
9	homogenous reproducible query into quality I
10	guess is one way to put it.
11	For whatever the explanations are
12	I think we can be at least somewhat comforted
13	by the notion that the reliability, the
14	signal-to-noise assessment of this specified
15	measure remains as good or better than
16	anything else we see down to levels of 30
17	cases assessed and perhaps below.
18	So, I personally take that as some
19	comfort and reassurance around the
20	specification of the measure.
21	CO-CHAIR KAPLAN: Thank you.
22	DR. JACOBS: This is Jeff Jacobs.

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	Page 54
1	I agree with everything you just said and just
2	to provide some clinical context.
3	A hospital doing 30 cases a year
4	of coronary artery bypass grafting is a really
5	low-volume hospital. I mean that's
6	CO-CHAIR HALL: Jeff, this is 30
7	cases over three years.
8	DR. JACOBS: Yes, I'm getting to
9	that. So 30 cases a year since you're doing
10	about 2 a month, a little over 2 a month.
11	Thirty cases over three years means that it's
12	one of the most low-volume hospitals on the
13	planet. It's not where I would go for my
14	coronary artery bypass graft.
15	CO-CHAIR KAPLAN: I think we're
16	getting the drift. Larry, can you make a
17	concise comment?
18	DR. GLANCE: Always.
19	(Laughter)
20	DR. GLANCE: So, this is very
21	concise. I think the point that Karen makes
22	is a really important one. And it is

Page 55 1 crosscutting. And the point is, and it will need 2 to be looked at at some point because it's 3 applicable to all the measures. 4 When you use shrinkage estimators 5 what you end up doing is classifying virtually 6 all of the low-volume providers as if they 7 were average and grouping them together with 8 other higher-volume centers that may in fact 9 10 be average. So that is a problem and I think 11 it will need to be addressed at some point. CO-CHAIR KAPLAN: Absolutely. 12 13 There is absolutely no disagreement that lowvolume hospitals remain a perplexing and 14 problematic issues for all of these outcome 15 16 And we probably aren't going to measures. 17 resolve that here now. On the other hand, for this 18 measure it looks like the low-volume issue is 19 20 probably as not problematic as we're going to 21 get. So, having said that, any other 22

Page 56 1 comments or questions? Okay, are we ready to vote usability? Go. 2 MS. SHAHAB: Voting for usability 3 and use, 1 high, 2 moderate, 3 low, 4 4 5 insufficient information. And your time 6 begins now. One more vote, please. We have all the votes for 7 8 usability and use. Thirteen high, nine 9 moderate, zero low, zero insufficient 10 information. 11 CO-CHAIR KAPLAN: Thank you. So now we're onto endorsement. 12 13 CO-CHAIR HALL: I have no additional concerns. 14 15 CO-CHAIR KAPLAN: Other comments 16 or questions? Larry, did you have your gizmo 17 up? Okay. Other comments? Are we ready to 18 go? Voting. MS. SHAHAB: Voting for overall 19 20 suitability for endorsement, 1 yes, 2 no. 21 Time begins now. All votes are in for overall 22

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	Page 57
1	suitability for endorsement. For measure 2514
2	Risk-adjusted Coronary Artery Bypass Graft
3	Readmission Rate, 22 yes, zero no.
4	CO-CHAIR HALL: We thank our
5	developers for their input. I guess we
6	neglected to ask whether Dr. Shahian or Jane
7	Han had anything to add, but too late for
8	those of you on the phone. We thank you for
9	your input.
10	And we'd like to ask the next
11	developers, Yale, CMS to come to the table.
12	Thank you. We have our developers for the
13	next measure 2515 to the table. Lein, Lisa
14	and Elizabeth are with us.
15	So if you wouldn't mind briefly
16	introducing yourselves and then your measure.
17	DR. SUTER: My name is Lisa Suter.
18	I'm from the Yale Center for Outcomes,
19	Research and Evaluation. And we're
20	introducing the measure 2514.
21	CO-CHAIR KAPLAN: Lisa, can you
22	move just a tad closer to the mike? Thanks.

	Page 58
1	DR. SUTER: Sure. Can you hear me
2	now? Great.
3	So I think the overall
4	CO-CHAIR HALL: Lisa, I'm sorry.
5	Did you say 2514? 2515?
6	DR. SUTER: 2515. My apologies.
7	CO-CHAIR HALL: Fifteen, thank
8	you.
9	DR. SUTER: I think the discussion
10	with the STS measure very nicely summarized
11	the collaborative process which CMS allowed us
12	to engage in with STS to develop these two
13	measures.
14	They are as harmonized as two
15	measures I think could possibly be. And the
16	success with which the claims-based measure
17	was able to achieve cohort and risk adjustment
18	validation is certainly due to the close
19	collaboration that we were able to participate
20	with STS's surgeons and their workgroup.
21	And I think we've talked about how
22	important CABG is as a readmission measure so

	Page 59
1	I'm not going to speak to that individually.
2	This measure differs slightly from
3	the registry-based measure in that it measures
4	all-cause unplanned readmissions after
5	isolated CABG procedures. And similarly to
6	use as a vetted guideline concordant approach
7	to measure development that's been supported
8	by the MAP.
9	As I mentioned the measure was
10	developed in close collaboration with STS and
11	every step of the measure development was
12	performed in parallel with the STS measure
13	developers and clinical experts.
14	I think as Dr. Jacobs mentioned
15	both measure developers recognized that there
16	are pros and cons to each measure, and that
17	they each have a place in the measurement
18	process.
19	The registry measure noting, as he
20	said, a clinical-based risk adjustment model
21	that I think has a greater face validity among
22	clinicians but a lower penetrance in terms of

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	Page 60
1	the number of hospitals that can be captured
2	and assessed without burden upon the
3	hospitals.
4	And as noted before we have
5	reached over a 97 percent agreement in the
6	cohort definition with the only discrepancies
7	either being distinct measure decisions such
8	as MAZE procedures which were made in concert
9	with the measure developers, or with
10	incongruities that can be ascribed neither to
11	the claims nor to the registry data
12	specifically and could represent errors in
13	either data source.
14	I think the other clarification
15	I'd just like to offer, and I'm sure we'll
16	have other discussions as we move on, is that
17	while this measure was developed in the over-
18	65 population I think it was noted that we
19	have assessed it in a California all-payer
20	data source.
21	This is to allow flexibility for
22	other users of this measure to use it in an

	Page 61
1	all-payer data source. But it was in fact
2	developed in a full national Medicare over-65
3	population data set.
4	And finally, just to remind people
5	that for other readmission measures that CMS
6	has implemented regarding the low-volume
7	hospital discussion there has been always
8	been an opportunity for hospitals to be noted
9	either that they are no different from
10	average, or that they are too small-volume to
11	be ascribed to a particular category. So that
12	that inability to categorize due to small
13	sample size is transparent to users. Thank
14	you very much.
15	CO-CHAIR HALL: Thank you, Lisa.
16	Lein, any comments or anything else to add?
17	DR. HAN: Hi. I am Lein Han from
18	CMS. And I just want to say that we
19	appreciate very much the collaboration with
20	STS. The working relationship was very good
21	and I really appreciate that. So thank you.
22	CO-CHAIR HALL: Thank you. So

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	Page 62
1	we're in the category of evidence. I would
2	like to turn to our primary discussants Ross
3	and John and ask them to open the discussion.
4	DR. EDMUNDSON: Yes, this is
5	ground that we covered here. But the evidence
6	is pretty compelling that there is a
7	readmission problem in this population and
8	that there's opportunity for improvement in
9	that. So I think that's well established.
10	DR. BULGER: I don't have anything
11	to add. From an evidence standpoint it's very
12	similar to the last measure which we looked
13	at.
14	CO-CHAIR HALL: I'm not seeing any
15	other cards so we'll vote evidence.
16	MS. SHAHAB: Voting for 1(a)
17	evidence, 1 yes, 2 no. Time begins now. Just
18	one more vote.
19	We have all the votes for 1(a)
20	evidence. Twenty-two yes, zero no.
21	CO-CHAIR HALL: Performance gap.
22	Any additional commentary above and beyond

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	Page 63
1	what we've reviewed? John?
2	DR. EDMUNDSON: Just on the
3	information provided that they have a mean of
4	16.8 percent readmission in the range of 12 to
5	22.1 percent. But again it's information
6	that's redundant as to what we discussed
7	before.
8	CO-CHAIR HALL: Wes?
9	DR. FIELDS: Yes, a question you
10	can either answer now as developers or later
11	in the process if you think it's more
12	appropriate.
13	But I'm just curious if analysis
14	of claims data reveals whether there is a
15	greater degree of variation or less than
16	optimal outcomes among the sites that don't
17	participate in STS or not.
18	So I'm just curious about whether
19	or not that's been part of your analysis and
20	whether you think there may be greater
21	variation in those few remaining sites that
22	aren't part of the STS registry.

	Page 64
1	DR. SUTER: This is Lisa Suter
2	from Yale. It's an excellent question.
3	Because of the proprietary nature of the STS
4	data we do not have the ability we at Yale
5	do not have the ability to identify individual
6	hospitals that either matched or did not
7	match.
8	I can't actually speak to whether
9	or not STS has investigated that among the
10	hospitals that they were unaware that did not
11	match to the Medicare data.
12	I know that many of the hospitals
13	that were not included in the validation
14	process because they did not link or match did
15	not have active participation in the STS
16	registry which represents about 10 percent of
17	hospitals in the nation.
18	DR. FIELDS: But do you know any
19	ballpark on the number of total CABGs in
20	Medicare that did not appear to have an STS
21	case match?
22	DR. SUTER: So, speaking to the

	Page 65
1	isolated CABG cohort that we identified,
2	approximately 30,000 patients or one-fifth of
3	the national isolated CABG patients identified
4	in claims approximately did not link or match.
5	DR. FIELDS: To an STS case.
6	DR. SUTER: To the STS. I don't
7	know the outcome rate among we did not
8	investigate the outcome rate among those
9	patients.
10	CO-CHAIR HALL: Wes, does that
11	answer your question or is it as close as
12	we're going to get for now?
13	DR. FIELDS: Well, we'll probably
14	talk more about it. I just find it
15	fascinating. I mean if ultimately what we're
16	trying to do is to reduce the remaining
17	variation.
18	You know, I have a lot of respect,
19	regard for the STS process and registry and
20	measure, but it raises a question about which
21	measure is most likely to actually reduce
22	variation going forward.

Page 66 1 CO-CHAIR HALL: Great. Any other additional comments? I'm not seeing any cards 2 3 raised for performance gap. MS. SHAHAB: Voting for 1(b) 4 performance gap, 1 high, 2 moderate, 3 low, 4 5 insufficient. Time begins now. 6 We have all the votes for 1(b) 7 performance gap. Nine voted high, thirteen 8 voted moderate, zero low and zero 9 10 insufficient. 11 CO-CHAIR HALL: Priority, John? DR. BULGER: So this has a similar 12 13 priority to the last measure. And as noted before this is one of MedPAC's targeted 14 diagnoses for readmissions priority. 15 16 CO-CHAIR HALL: Okay, any 17 additional comments? Not seeing any. MS. SHAHAB: Voting for 1(c) high 18 priority. One is high, two moderate, three 19 low, four insufficient. Your time begins now. 20 We have all the votes for 1(c) 21 22 high priority. Eighteen voted high, four

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	Page 67
1	voted moderate, zero low and zero
2	insufficient.
3	CO-CHAIR HALL: Moving into the
4	scientific realm, reliability and validity.
5	John, Ross, you want to open the discussion?
6	DR. EDMUNDSON: Okay, reliability.
7	This is a test/retest split sample. And with
8	intraclass correlation coefficient here.
9	Large numbers.
10	What I found interesting and
11	they used Medicare claims for the years 2008,
12	09 and 10 as well as the Society of Thoracic
13	Surgeons. But I believe this is on the claims
14	data that you're doing the split samples, is
15	that correct?
16	DR. SUTER: That's correct.
17	DR. EDMUNDSON: And then on that
18	the random split samples for each hospital was
19	the intraclass correlation coefficient was
20	0.331 which was judged as fair. Could
21	developers comment on that relationship?
22	DR. SUTER: Yes, we heard during

	Page 68
1	the workgroup the concerns about the low
2	intraclass correlation coefficient during the
3	split sample retest.
4	So, trying to understand the
5	stability of the measure result, we think that
6	the most robust and conservative assessment is
7	to fully separate the sample of patients so
8	that in an individual hospital there is no
9	overlap between the two samples.
10	So we randomly split each
11	hospital's patients into equal portions and
12	then we calculate the risk-standardized
13	readmission rate at the hospital level in each
14	of those samples.
15	And using a 3-year data set which
16	was what was available to us we received we
17	yielded an ICC of 0.33 which you would agree
18	is outside of the range of 0.4 to 0.7 which is
19	usually interpreted as fair to good or
20	moderate for intraclass correlation
21	coefficients.
22	Although in many, most of the

Page 69 1 readmission measures other than the hospitalwide readmission measure that CMS has 2 3 implemented use three years of data in order to achieve a sample size. And this is 4 particularly true in a procedural-based 5 measure such as CABG. 6 In response to the concerns about 7 the ICC and based on some recommendation from 8 9 prior NQF discussions for other measures using 10 a method similar to John Adam's paper which 11 was referenced in regards to RAND's method using the Spearman-Brown prophecy formula --12 13 and we have this information available to the committee if you'd like to see it -- we 14 estimated what would the intraclass 15 correlation coefficient be if we were actually 16 17 able to create a 3-year sample that could be split into two equal 3-year samples. So that 18 you had the volume of the 3-year sample but 19 20 you still had two completely independent and 21 non-overlapping samples. 22 And when we perform that analysis

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	Page 70
1	for CABG the ICC rises to 0.5. It rises for
2	all of the readmission measures that our
3	measure developer has in front of the
4	committee today and we'd be happy to share
5	this information. We have copies that we can
6	provide to you.
7	CO-CHAIR HALL: Additional
8	comments on this area, reliability?
9	Reproducibility. Not seeing any go ahead.
10	CO-CHAIR KAPLAN: I would say that
11	for these intraclass correlation coefficients
12	this is roughly in the range of the things you
13	see.
14	And in part and we were just
15	having a dialogue about it. In part it's
16	because within hospitals when you're looking
17	across patients within hospitals you're
18	dealing with a dichotomous variable of
19	readmitted/not readmitted. So it's a little
20	bit of a compressed variance problem.
21	But having said that there are,
22	you know, this is in the zone where you

	Page 71
1	exactly what you would see when you look at
2	other kinds of measures we've already
3	considered for the intraclass correlation
4	coefficient.
5	CO-CHAIR HALL: Not seeing any
6	cards we'll vote on reliability.
7	MS. SHAHAB: Voting for 2(a)
8	reliability, 1 is high, 2 moderate, 3 low, 4
9	insufficient and your time begins now.
10	We have all the votes for 2(a)
11	reliability. One voted high, twenty-one
12	moderate, zero low and zero insufficient.
13	CO-CHAIR HALL: And validity?
14	Ross, John, opening comments.
15	DR. BULGER: So a couple of
16	questions on validity. In general, the C
17	statistic was 0.63 which is similar to the
18	last measure.
19	This is administrative data, not
20	clinical data so I wondered if you at some
21	point could speak to that.
22	The other question that came up

	Page 72
1	with the last data was with the LVAD patients.
2	And they are in this group. And we had just
3	talked about the ability to exclude one subset
4	of those.
5	But my assumption, and you can
6	speak to this, that because of the
7	administrative yours is administrative data
8	that you could only exclude them totally or
9	not exclude them, but not subset them into
10	elective and non-elective LVAD patients.
11	The face validity from the panel I
12	think was strong as well. In looking at your
13	exclusions they were similar to the last one
14	we looked at.
15	There was a question of excluding
16	patients from the panel in our discussions
17	from the workgroup of excluding patients who
18	died in the 30 days.
19	And I think you had mentioned, you
20	made some comments back on that already that
21	they were in because that was what was similar
22	to the there Yale measures, to keep those in.
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	Page 73
1	But I was wondering if you could speak to that
2	as well.
3	DR. SUTER: Great, thank you. And
4	I also just wanted to correct an answer. I
5	had responded to the question of capture. My
6	off-the-cuff math was off. So instead of
7	being 20 percent it's 10 percent non-capture
8	rate. I apologize for that error.
9	In regards to post-discharge
10	mortality. So patients who die within the
11	hospital obviously are not at risk for
12	readmission. They are in fact excluded from
13	the measure.
14	There are a small proportion of
15	patients who die after discharge from the
16	hospital. That is about 1.4 percent of
17	patients. They do as expected have a higher
18	readmission rate.
19	This measure is paired with a
20	mortality measure which is in front of the
21	surgery committee and will be reviewed by the
22	NQF later this year.

	Page 74
1	We think that that allows the
2	capture of a full spectrum of quality
3	outcomes. So to prevent any unintended
4	consequences of measuring readmission in this
5	cohort of patients.
6	In regards to the VAD issue, as
7	has been previously noted we do include VAD
8	procedures based on the recommendation of a
9	host of cardiothoracic surgeons who are
10	involved in this measure development. And it
11	is harmonized with the STS measure.
12	As you noted we do not have the
13	ability to finesse the the ability to
14	identify unplanned versus planned VAD
15	procedures. In a cohort of about 150,000
16	patients with isolated CABG only 90 have VAD
17	procedures.
18	About 50 percent of them have
19	percutaneous VAD procedures and they have a
20	readmission rate very close to the mean 17.5
21	where the average hospital readmission rate is
22	16.5 in non-VAD patients.

	Page 75
1	Those that have more invasive VAD
2	procedures do have a higher readmission rate
3	around 30 percent.
4	Because they represent 0.03
5	percent of the entire cohort of isolated CABGs
6	and they are not clustered in any one
7	particular hospital or type of hospital we as
8	measure developers are open to bringing back
9	to our workgroup the recommendation from NQF
10	to remove these procedures if there's a strong
11	feeling that they misrepresent the quality of
12	hospital performance using this measure.
13	We went ahead with the best
14	recommendations from the largest group of
15	cardiothoracic surgeons so we felt confident
16	in that recommendation and we are eager for
17	the NQF's advice.
18	And I will also talk about risk
19	adjustment. I don't know if people wanted to
20	comment on VADs before I move onto risk
21	adjustment.
22	So, in the materials that we

	Page 76
1	submitted we also submitted a technical
2	report. And on page 81 and 82 there are two
3	graphs that I think are particularly helpful
4	to understand the risk adjustment validation
5	that was performed with the Society of
6	Thoracic Surgeons.
7	And this is in a matched cohort of
8	patients as was previously discussed.
9	And I know that it's not easy to
10	see this, but these are all of the hospitals
11	that were identified as outliers among all of
12	the thousand or so hospitals included in the
13	validation process.
14	And each pair of lines and dots
15	represent the risk-standardized readmission
16	rate achieved with the claims-based measure
17	and that achieved with the registry-based
18	measure.
19	And while it's I know impossible
20	for you to see across a room, the gray area
21	represents the top of the gray area
22	represents the national rate.

	Page 77
1	And what I hope is visually
2	apparent is that those paired lines are
3	extremely overlapping. So in addition to
4	achieving an ICC of 0.9 something depending on
5	what ICC method you use so you can see that
6	the risk-standardized rate is highly
7	correlated, the interval estimates, the
8	uncertainty around that interval estimate,
9	excuse me, around that risk-standardized
10	readmission rate is also highly correlated.
11	And any time you draw a line to
12	move patients or hospitals into a performance
13	category which is what the nationally reported
14	readmission measures do. They report them as
15	better than average, worse than average, or no
16	different than average, or too small to
17	quantify. You have to draw a line.
18	And I'm happy to share this. It's
19	also in your materials. But when you draw
20	that line some people fall on one or the other
21	side.
22	But I think what's very reassuring

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	Page 78
1	about this is that the two measures produce
2	performance estimates for each hospital that
3	are qualitatively highly similar.
4	Quantitatively they do miscategorize a few
5	percentages of patients, but the specificity
6	in the claims-based measures is close to 100
7	percent. It's over 99 percent which I think
8	for a high-stakes measure which readmission
9	may be we think is the proper emphasis in
10	terms of being conservative.
11	CO-CHAIR HALL: Okay. I'm sorry,
12	I myself am trying to follow what you said
13	while looking at the diagram. So I'm slightly
14	lost on that diagram. But I know Larry has a
15	question.
16	DR. GLANCE: So, I just wanted to
17	comment on the validation part. And I want to
18	preface my comments by saying I understand the
19	need for a measure based on administrative
20	data because some of the hospitals in the U.S.
21	are not part of STS.
22	Having said that, you're

	Page 79
1	absolutely correct in that there's a very high
2	level of agreement between your measure and
3	the STS measure. In fact, according to your
4	reporting there's about 97 percent agreement
5	between the administrative data and the STS
6	measure in terms of classification as high
7	quality, average quality and low quality.
8	Having said that, when you look at
9	the sensitivity of the CMS measure for
10	identifying high-quality and low-quality
11	hospitals it's about 40 percent and 60 percent
12	respectively. So, I'd like you to comment on
13	that.
14	DR. SUTER: So, I think the
15	challenge with this validation process is in
16	this case we were validating the risk
17	adjustment. So we made the assumption that
18	the registry data represents the gold
19	standard.
20	We don't actually know what the
21	gold standard for performance categorization
22	is in the United States for isolated CABG

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	Page 80
1	procedure readmission rates. So, I can't
2	really comment on whether or not sensitivity
3	of 40 or 60 percent is appropriate or not.
4	I think it is a challenge when you
5	are creating measures that you want to be
6	responsive to change and useful to hospitals,
7	and yet they're being publicly reported and
8	you want the estimates to be stable and
9	reliable.
10	And in that situation we often
11	need longer measurement periods, larger data
12	sample sizes and we favor in the claims-based
13	measure and the registry measure uses the same
14	hierarchical modeling that does pull people
15	towards a less outlier position. But I think
16	we felt in the situation of how these measures
17	may be used that that was a reasonable
18	tradeoff.
19	And certainly the policy of
20	implementation is not our decision, but we
21	work in close concert with CMS to make sure
22	that we're responsive to their needs.

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	Page 81
1	CO-CHAIR KAPLAN: I'd like to just
2	sort of clarify some issues because from a
3	measurement person's perspective criterion
4	validity says there is a gold standard and
5	you're comparing a new measure to that gold
6	standard. There is no gold standard.
7	Convergent validity says I have
8	two sources of information and they're telling
9	me roughly the same thing. That's confidence-
10	inspiring from the graph I'm staring at right
11	now. Two data sources tell you roughly the
12	same thing.
13	Discriminate validity, however,
14	says I can tell hospitals apart. And from
15	that perspective not so much. So, can you
16	help us understand sort of in those terms?
17	So, criterion validity is off the
18	shelf. Convergent validity we're seeing
19	evidence of. What happens to discriminate
20	validity and can you tell hospitals apart?
21	DR. SUTER: So, we were unable to
22	de-identify the individual hospitals that are

	Page 82
1	discordant. So, it's challenging to dig into
2	the discordant hospitals to try and understand
3	why a particular hospital might have been
4	considered higher quality or lower quality
5	from one data source versus another.
6	I think it is an important
7	question. Unfortunately I can't speak to it.
8	CO-CHAIR KAPLAN: Thank you.
9	CO-CHAIR HALL: Any other
10	comments, concerns on the topic category of
11	validity? Wes?
12	DR. FIELDS: Yes, I just want to
13	come at this from a different direction. So
14	it's sort of reciprocal to my earlier comment.
15	I would assume that if you have a
16	clinical data set as the registry does that
17	you'd have more independent variables that
18	help you get at the nature of quality if you
19	will and to distinguish between facilities and
20	programs.
21	So I just want to ask a first
22	order question. Compared to the CMS claims

	Page 83
1	data set in terms of numbers of data points
2	how much larger is the data set the STS
3	registry uses compared to the number of
4	elements in a claim stream that CMS would
5	receive?
6	CO-CHAIR HALL: I want to push on
7	you, Wes. I can see this might shed some
8	light but we're only considering the measure
9	in front of us, right? We're not really
10	considering its comparison to an STS
11	counterpart.
12	So, if your question helps us get
13	to this measure I think we're okay. So, if
14	the developers can comment on it in that
15	light, in that context. Or did I
16	misunderstand, Wes?
17	DR. FIELDS: I thought I was
18	restating Sherrie's question from a different
19	context. I think the issue of how you define
20	quality is pretty interesting. And I'm
21	assuming that having more data elements from
22	a clinical registry that's larger in scope

	Page 84
1	than a claim stream gives you the possibility
2	of doing that.
3	I'm just asking them to quantify
4	how many more variables are in the registry
5	stream. Because so it's really a way of
6	restating Sherrie's question about how you get
7	at the nature of quality and distinguishing
8	between facilities and programs.
9	DR. SUTER: So I think thank
10	you, Lisa Suter. The response is two that are
11	C statistics so our discriminate ability is
12	essentially identical.
13	And I think the other is that both
14	measure I mean, certainly our we as the
15	measure developer see room for both of these
16	measures in the world of measurement. They
17	offer unique perspectives. They were
18	developed in an incredibly harmonized fashion
19	and offer advantageous synergistic information
20	about hospital performance, not necessarily
21	replacement performance.
22	CO-CHAIR HALL: And they do

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	Page 85
1	provide in their methodology report these
2	insights such as you're seeing on this graph
3	to try to help us understand whether one
4	approach or the other approach helps
5	discriminate the quality better.
6	The underlying notion that more
7	variables will help you discriminate quality
8	better may or may not be true. More variables
9	could lead you to decide there is no
10	difference in quality. So, the underlying
11	construct is still ill-defined or
12	controversial.
13	Other comments? Wes, do you have
14	more comments? Any other comments in this
15	category?
16	CO-CHAIR KAPLAN: I'd just add one
17	thing. More variables usually help you
18	improve your estimates of reliability, not
19	necessarily your estimate of validity. So,
20	reliability you ask more things, you get a
21	tighter, more reliable response. But not
22	necessarily a more valid response.

Page 86 1 CO-CHAIR HALL: Okay, I'm not seeing any cards raised so we'll move on 2 3 validity. MS. SHAHAB: Voting for 2(b) 4 validity. One is high, two moderate, three 5 low, four insufficient and your time begins 6 7 now. We have all the votes for 2(b) 8 9 validity. Two high, twenty moderate, zero 10 low, zero insufficient. 11 CO-CHAIR HALL: Feasibility? Ross? 12 13 DR. EDMUNDSON: Feasibility, I think this is claims data. This is very 14 feasible. We can do this. 15 16 CO-CHAIR HALL: Any other 17 concerns? I don't see any raised. MS. SHAHAB: Voting for criteria 3 18 feasibility. One is high, two moderate, three 19 20 low, four insufficient and your time begins 21 now. We have all the votes for 22

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	Page 87
1	feasibility. Twenty voted high, two moderate,
2	zero low and zero insufficient.
3	CO-CHAIR HALL: Usability. John?
4	DR. BULGER: So I think the
5	usability is similar at least to the last
6	measure. I think the question came up in our
7	pre-work is, you know, this discrimination
8	issue of high performers, mid performers and
9	low performers. And if this goes like some of
10	the other measures that have been used which
11	are set at a midpoint really from a payment
12	standpoint I think there was concern amongst
13	the group how that would perform.
14	And the other concern was if it
15	really only was able to discriminate the tails
16	what the use to the public would be from that
17	standpoint.
18	Otherwise, I think we were fine
19	with what we were looking at.
20	CO-CHAIR HALL: Paul?
21	DR. HEIDENREICH: Yes, just in
22	terms of going forward from CMS' perspective.

	Page 88
1	I think there's going to be 20 there's
2	about 20 hospitals rated as good or better
3	than expected. Is that about right? Which
4	some people have said is very low.
5	I'm not sure what the right
6	percentage we should label as outliers, but is
7	there any plan to change that implementation
8	when this is reported on the website? About
9	which one what fraction are outliers versus
10	what fraction are not outliers?
11	DR. SUTER: So from a measure
12	developer standpoint this is Lisa Suter
13	that you're referring to the 2008-2010 data.
14	I can't speak to more recent data. I don't
15	have those estimates in front of me.
16	Presumably more recent data would be used for
17	reporting purposes.
18	And in terms of whether or not to
19	use the same performance categorizations that
20	are used in other measures I'll defer to Dr.
21	Han.
22	DR. HAN: Hi, this is Lein Han

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	Page 89
1	from CMS. At this moment we plan to continue
2	the way we display the data on Hospital
3	Compare.
4	And I believe that you all know
5	but I just want to describe again the four
6	categories. We have three compared to the
7	national rate that's better, worse, or no
8	different. And the other one is the category
9	of small hospital that have less than 25, we
10	put them aside. So, that would be the way
11	this moment yes, that's the way we're going
12	to plan to display the measure. Thank you.
13	DR. SUTER: And I'll just add that
14	currently the interval estimates that were
15	used for the graphic that was up and are used
16	for the other measures reported on Hospital
17	Compare uses a 95 percent interval estimate.
18	If you felt, if the nation felt
19	that a larger number of outliers was a more
20	revealing information you could certainly
21	change the interval estimate for reporting
22	purposes in order to identify more outliers.

	Page 90
1	DR. HAN: We welcome suggestions
2	if you have any ideas.
3	CO-CHAIR HALL: And just to
4	clarify it looks like in the materials
5	submitted that there were about 1.2 percent
6	high/good outliers and 1.5 percent low/bad
7	outliers. So we're in the 1.2 to 1.5 percent
8	of institutions being labeled according to the
9	specifications you just heard.
10	DR. SUTER: And may I also add
11	that each hospital receives a hospital-
12	specific report for the currently publicly
13	reported readmission measures. This is Lisa
14	Suter.
15	And in that report they receive
16	detailed information about all of their
17	patients in the measures.
18	So, while there's a lot of focus
19	on who's an outlier on Hospital Compare, there
20	is still a tremendous amount of detailed
21	information reported back to hospitals.
22	And I know STS I'm sure has

Page 91 1 similar reporting back to their hospitals, detailed information about patients in the 2 cohort, who was included, who was readmitted. 3 This is information that's not available to 4 hospitals because a large proportion of 5 patients are readmitted to hospitals who did 6 not perform the CABG. So this kind of 7 information is incredibly valuable for quality 8 9 improvement purposes even if there is not a 10 distinct large number of outliers on a 11 publicly reported website. DR. HAN: Hi, this is Lein Han 12 13 from CMS. I just want to add to that. We also, CMS also offer the Q&A 14 It's like hospital when they get the 15 service. 16 data they can call -- they can email CMS any 17 question they have. Yes, I don't know whose phone 18 number I'll provide for who to call --19 20 (Laughter) DR. HAN: But we do have this 21 So, hospital can contact us any time 22 service.

	Page 92
1	they want.
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2	CO-CHAIR HALL: Larry?
3	DR. GLANCE: So, I have a comment
4	on usability. I think we both recognize that
5	we have two very, very high-quality measures
6	but they're essentially looking at the same
7	thing.
8	And the issue that I see as a
9	potential issue is that although all of us
10	understand in this room that risk adjustment
11	isn't perfect and that depending on which risk
12	adjustment model you may end up coming to
13	different conclusions, I'm not sure that all
14	the consumers of this information will
15	understand that.
16	And I think it goes to the heart
17	of credibility of performance measurement when
18	you potentially release in the public domain
19	two different report cards which significantly
20	disagree on which hospitals are identified as
21	high quality and which ones are identified as
22	low quality.

1	
	Page 93
1	And I would urge caution to the
2	developers, CMS and STS, that when you decide
3	which quality measures to release to the
4	public that maybe you agree on releasing one
5	set of measures as opposed to both for the
6	hospitals where you have overlap.
7	DR. HAN: Hi, this is Lein Han,
8	CMS. We think these CABG is a very important
9	area. We all recognize that. And we think
10	that these two models are pretty good. Two
11	very good measures.
12	I think the consideration for CMS
13	is what is the most cost-effective and less
14	burdensome way for hospitals to implement the
15	measures. And that's really our consideration
16	right now.
17	For us claims is the most cost-
18	effective way to implement input in measure.
19	So, that would be our priority to have a way
20	to implement most feasible to both CMS and to
21	hospitals.
22	CO-CHAIR HALL: Any additional

	Page 94
1	concerns? Thoughts?
2	We know that the claims-based
3	measure has been written into IPPS for 2017
4	and the STS one has not. That's my
5	understanding. I don't think that affects our
6	decision about whether the measure in front of
7	us is useful but we I think it does relate
8	to Larry's comment and in fact relates to
9	Wes's as well. Larry's and Wes's comments
10	both related to comparison between two
11	measurement programs that could give some
12	different results.
13	In fact, we know the registry-
14	based program seems to identify at least twice
15	as many institutions on the tails as the
16	claims-based program. So there's a danger for
17	what Larry's concerned about.
18	But in this case the plan is to
19	implement. We know this measure is written
20	into IPPS for 2017 and again, that I don't
21	think affects the overall assessment of this
22	measure or should affect it.

Page 95 1 Any other concerns or questions before we vote usability? I don't see any. 2 MS. SHAHAB: Voting for usability 3 and use, 1 high, 2 moderate, 3 low, 4 4 insufficient information. Your time begins 5 6 now. 7 We have all the votes for usability and use. Three high, eighteen 8 9 moderate, one low and zero insufficient 10 information. 11 CO-CHAIR HALL: Okay, before we vote overall any additional concerns or 12 13 comments? Summary comments? Wes, you're just 14 smiling. 15 DR. FIELDS: Bruce, I'm just so 16 happy to be here to participate in the 17 process. Thank you so much. 18 (Laughter) CO-CHAIR HALL: Why do I feel like 19 20 Wes is coming after me. Any final comments? 21 Not seeing any. MS. SHAHAB: Voting for overall 22

1	
	Page 96
1	suitability for endorsement, 1 yes, 2 no.
2	Time begins now.
3	We have all the votes for overall
4	suitability for endorsement for measure 2515
5	Hospital 30-day All-cause Unplanned Risk-
6	standardized Readmission Rate Following
7	Coronary Artery Bypass Graft Surgery, 21 yes,
8	1 no.
9	MR. AMIN: So, before we move onto
10	the next measure I just want to remind the
11	committee and the developers that these
12	recommendations for endorsement are still
13	contingent on a conversation related to
14	competing measures.
15	So this measure and the STS
16	measure will be discussed in terms of how
17	they're, you know, whether they're competing
18	and whether it's justified to have both
19	measures in the portfolio.
20	And that is in addition to this
21	SNF measure, SNF readmission measures that we
22	discussed yesterday. We likely won't have

	Page 97
1	time for that discussion during today's in-
2	person meeting but we will have a follow-up
3	call to discuss that.
4	CO-CHAIR HALL: I've been trying
5	to resist going down that road but your
6	comment makes it irresistible to ask this
7	question though.
8	We just talked about two measures
9	but we know one of them is written into IPPS.
10	So does that not affect that discussion around
11	competition between those measures?
12	DR. BURSTIN: You know, it's a
13	great question, Bruce. I personally feel like
14	for this committee's sake it's really about
15	the comparability of the measures themselves
16	and about the questions of whether you can in
17	fact from purely a perspective of use broadly
18	have both of those measures out there. Will
19	it add to confusion? Can people understand
20	the nuances? How much does the difference in
21	data source affect the way people may use
22	them? So I don't think it has a particular

Page 98 1 issue. I think it may very well come up 2 at the MAP certainly where they are 3 specifically charged with looking at which 4 measures for which programs. But I don't 5 think it should particularly have an impact on 6 the discussion around competing -- and in this 7 instance it's not really harmonization. 8 They 9 are fully harmonized except for data source. 10 I think it is still a competing 11 issue and I think Larry's comments really raised that issue significantly in terms of 12 13 understanding comparability, the comments related about the 15 percent of people who 14 aren't in STS who are in this measure. 15 16 I mean there's just many issues I 17 think you'll have a chance to chew on I assume in a separate conference call to follow. 18 CO-CHAIR HALL: We thank the 19 20 developers from CMS, Yale. Are the same folks going to stay 21 at the table for the next? Or will it be a 22

1	
	Page 99
1	different team? New crew?
2	CO-CHAIR KAPLAN: Okay, this is
3	measure number 2513 Hospital 30-day All-cause
4	Risk-standardized Readmission Rate Following
5	Vascular Procedures. The developer is Yale.
6	Could you please briefly introduce
7	yourselves and then the measure. Is anyone on
8	the phone? No, everyone is here in the room.
9	Excellent.
10	DR. MCNAMARA: Hi, I'm Bob
11	McNamara. I'm a cardiologist at Yale. Jeptha
12	Curtis is next to me, another cardiologist at
13	Yale. Susannah Bernheim, also on the team is
14	behind us here and we have multiple people on
15	the phone including Lori Geary who's a part of
16	the team.
17	I understand you have the whole
18	measure in front of you. I just wanted to
19	have a few to give a few highlights
20	regarding this measure.
21	It's a very important measure.
22	Vascular surgery and readmission was

	Page 100
1	identified in the MedPAC report as one of the
2	seven conditions that were responsible for up
3	to 30 percent of the preventable readmissions.
4	The high cost, high readmission, high
5	variation right along with the information
6	that providers and hospitals and physicians
7	need for quality development and patients need
8	for choice. So that was the first one.
9	The second highlight is going into
10	this measure we knew it was going to be very
11	complex. We knew we would need a lot of
12	clinical input both on our team and within our
13	technical expert panel, the technical expert
14	panel which was highly competent and engaged
15	in the whole process involving multiple
16	different specialties that are going to be
17	affected by this measure, vascular surgeons,
18	interventional radiologists, interventional
19	cardiologists as well as experts in
20	methodology, policy and patient advocate. And
21	they were involved from the beginning for many
22	if not all of the major decisions.

Page 101 1 A third also regards the complexity. As opposed to some of the other 2 measures this is going to have many different 3 procedures. So, identification of the 4 5 procedures and ultimately the patients was 6 going to be very critical. We developed a few guiding 7 principles right from the beginning to 8 9 identify which procedure should be included. 10 First, it was going to be a major 11 procedure that was going to be involved. We didn't want to include venal punctures, 12 arterial catheterizations and things like 13 14 that. 15 It had to be clinically coherent. 16 Initially MedPAC called it other vascular 17 meaning didn't want cardiac, didn't want intracranial. 18 We also made the decision not to 19 20 include hemodialysis catheter-related thrombectomies and the like. 21 And the third criteria was it had 22

	Page 102
1	to be central to the hospitalization. We
2	didn't want the vascular surgery to be a
3	suturing of an artery from another surgery.
4	So those were some of the guiding principles.
5	Finally, another major highlight
6	regards the risk adjustment that we used. The
7	typical hierarchical model including both the
8	patient characteristics as well as clustering
9	of patients within a hospital.
10	In addition to the patient
11	characteristics we wanted to include the
12	different procedures. So we grouped the
13	procedures in eight different categories.
14	They included both anatomical location at
15	neck, thoracic, abdominal and limb as well as
16	an unspecified.
17	And we wanted to be inclusive as
18	possible, include both endovascular procedures
19	and open. So there's many other aspects of it
20	but just wanted to give you those highlights.
21	And open for any questions. Thank you.
22	CO-CHAIR HALL: So we apologize at

Page 103 1 the table. We've been doing a little bit of whispering while Robert was talking. 2 We 3 apologize for that. I was an expert on this measure 4 for Yale and so I'm going to recuse myself 5 from this discussion and that's what we've 6 been whispering about. So I'll turn over to 7 Sherrie. 8 9 CO-CHAIR KAPLAN: Okay. So, I did 10 not review this measure and the other 11 reviewer, Paulette, is not also with us today. So I am going to be looking at Bruce's --12 13 MR. AMIN: There are a number of 14 workgroup members --CO-CHAIR KAPLAN: Who were on the 15 16 workgroup. 17 MR. AMIN: Yes. CO-CHAIR KAPLAN: So I'm going to 18 look at Bruce's notes as best I can and the 19 20 rely -- who was on the workgroup? Hands? 21 Okay, so at least some people here have -- I did not review this measure so I will look at 22

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	Page 104
1	Bruce's notes as best I can. And then we will
2	count on the workgroup members to kind of
3	pitch in here.
4	Okay, so with respect to the
5	evidence, comments from the workgroup?
6	MS. KHAN: So the workgroup 1
7	members were John Bulger, Bruce Hall, Mae
8	Centeno, Ross, Paul, Larry, Cristie and
9	Paulette.
10	CO-CHAIR KAPLAN: Comments?
11	Larry?
12	DR. GLANCE: The evidence is very
13	strong for this measure.
14	CO-CHAIR KAPLAN: Other comments
15	from anybody else on the workgroup? Okay, I
16	guess we're ready to vote.
17	MS. SHAHAB: Voting for 1(a)
18	evidence. One is yes, two is no and your time
19	starts now. We still need two more votes.
20	We have all the votes for 1(a)
21	evidence. Twenty yes, one no.
22	CO-CHAIR KAPLAN: Okay.

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	Page 105
1	Performance gap. Larry, do you want to speak
2	to that?
3	DR. GLANCE: So there's good
4	evidence of a performance gap. Between the
5	10th percentile and the 90th percentile the
6	risk-standardized readmission rates were 12.3
7	percent versus 14.9 percent respectively. So,
8	about an over 2.5 percent absolute difference.
9	CO-CHAIR KAPLAN: Yes, the
10	interquartile range was 12.9 to 14.3. So, but
11	the range looks like 10 to 18 percent from the
12	highest to the lowest. So there appears to me
13	as well to be compared to some of the other
14	measures we've seen a performance gap. Paul?
15	DR. HEIDENREICH: I agree there's
16	a gap. It doesn't seem as large to me as some
17	of the other gaps. But that still leaves
18	moderate.
19	CO-CHAIR KAPLAN: So I think
20	perspective is everything. I saw one where
21	the interquartile range was 0.9 percent. So
22	it kind of depends. But at least it's in the

Page 106 1 range I think of the ones that we've seen. Any other comments? Ready to vote performance 2 3 gap? MS. SHAHAB: Voting or 1(b) 4 performance gap. One high, two moderate, 5 three low, four insufficient. Time begins 6 7 now. We have all the vote for 1(b) 8 9 performance gap. Four high, seventeen moderate, zero low, zero insufficient. 10 11 CO-CHAIR KAPLAN: Priority. Larry, do you want to -- anybody from the 12 13 workgroup want to say anything? DR. FIELDS: I'd just say this 14 falls again -- this is one of MedPAC's seven 15 conditions which account for 30 percent of all 16 readmissions in the Medicare program. So it's 17 18 a high priority. CO-CHAIR KAPLAN: Thank you. 19 20 Anyone else? Voting priority. 21 MS. SHAHAB: Voting for 1(c) high 22 priority. One is high, two moderate, three

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	Page 107
1	low, four insufficient. Time begins now. One
2	more vote.
3	We have all the votes for 1(c)
4	high priority. Sixteen voted high, five voted
5	moderate, zero low and zero insufficient.
6	CO-CHAIR KAPLAN: Scientific
7	acceptability. First up is reliability.
8	Larry, do you have comments?
9	DR. GLANCE: So this is 30-day
10	all-cause unplanned readmissions. This was
11	done using hierarchical modeling as per the
12	standard approach.
13	They adjusted for age, sex,
14	demographics, procedures and clinical
15	covariates using the hierarchical condition
16	categories, a fairly standard approach.
17	In terms of reliability testing
18	the standard approach yielded an intraclass
19	correlation coefficient of 0.4 which is very
20	much in the zone, maybe in the upper level of
21	that zone.
22	CO-CHAIR KAPLAN: Thank you.

	Page 108
1	Others? Are we ready to vote reliability?
2	Any other comments? Okay.
3	MS. SHAHAB: Voting for 2(a)
4	reliability. One is high, two moderate, three
5	low, four insufficient and the time begins
6	now.
7	We have all the votes for 2(a)
8	reliability. Two high, nineteen moderate,
9	zero low and zero insufficient.
10	CO-CHAIR KAPLAN: Now, validity.
11	Larry?
12	DR. GLANCE: So, the measure
13	developers convened a technical expert panel
14	who expressed strong support for the face
15	validity of this measure.
16	They validated this model in an
17	independent data set. It had a C statistic of
18	0.67 which is at the upper end of the zone of
19	acceptability for these readmission measures.
20	They looked at calibrations both
21	graphically and also using a standard
22	methodology and the model was well calibrated,
	Page 109
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1	showed goodness of fit.
2	CO-CHAIR KAPLAN: So, the
3	discriminate validity is sort of yet to be
4	determined. Is that correct? So we don't
5	have a sense of whether this discriminates
6	well between hospitals.
7	DR. GLANCE: I don't recall
8	exactly how many hospitals were labeled as
9	high-quality and low-quality. Maybe the
10	measure developers could address that?
11	DR. MCNAMARA: Yes, this is Bob
12	McNamara. We did not do that analysis feeling
13	that this is to develop the measure, the
14	implementation can be any cut point that you
15	want. Not to put it in more of a policy
16	decision.
17	We had talked about that with the
18	prior measure. We can address it now if the
19	committee wants to. I don't know if there's
20	much more to say on that.
21	CO-CHAIR KAPLAN: Any other
22	comments from the working group or the

Page 110 1 steering committee? Hearing none voting validity. 2 3 MS. SHAHAB: Voting for 2(b) validity. One is high, two moderate, three 4 low, four insufficient. Time begins now. 5 We have all the votes for 2(b) 6 validity. Zero voted high, twenty voted 7 moderate, zero low and one insufficient. 8 9 CO-CHAIR KAPLAN: Thank you. 10 Feasibility? 11 DR. GLANCE: So this is a highly feasible measure. It's based on widely 12 13 available administrative data. 14 CO-CHAIR KAPLAN: Any other comments or questions? Nope? Ready to vote 15 16 feasibility. 17 MS. SHAHAB: Voting for feasibility. One high, two moderate, three 18 low, four insufficient. Time begins now. 19 We have all the votes for 20 21 feasibility. Seventeen voted high, four voted moderate, zero low and zero insufficient. 22

	Page 111
1	CO-CHAIR KAPLAN: Thank you.
2	Usability and use. Larry, do you have any
3	comments?
4	DR. GLANCE: This one's a little
5	bit more difficult to comment on. It's a new
6	measure so we don't have too much information
7	on the usability of this particular measure.
8	CO-CHAIR KAPLAN: Kathy?
9	DR. AUGER: I think it's a little
10	challenging to assess usability and use if we
11	don't know how many outlier hospitals there
12	are. So we don't know whether it's able to
13	really discriminate high performers from low
14	performers. And so it just makes it
15	challenging for me to assess.
16	CO-CHAIR KAPLAN: Taroon or Helen,
17	you want to comment on when a measure is early
18	on in the phase of development how that works?
19	MS. PACE: Yes. So, basically
20	what we ask the developer to do is to do two
21	things when it's a new measure. To write up
22	how they think it can be used in improvement,

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	Page 112
1	how it will be used for improvement and what
2	are the plans for its use in accountability
3	applications.
4	So, we can look at that section of
5	their form or maybe the developers just want
6	to remind people what they indicated as far as
7	how this measure can accommodate improvement
8	as well as plan for accountability
9	applications.
10	CO-CHAIR KAPLAN: Developers?
11	DR. MCNAMARA: This is Bob
12	McNamara. I think we're going to let Lein
13	talk about that from CMS.
14	DR. HAN: This is Lein Han from
15	CMS. So, I think I have to give Yale the
16	developer, our contractor, credit. Because
17	for this measure to keep the integrity of the
18	measure they actually include cases from both
19	inpatient and outpatient settings. Am I
20	correct?
21	So, right now CMS is trying to
22	figure out which program, either IQR, it means

	Page 113
1	inpatient quality reporting program, or
2	outpatient quality reporting program, that
3	this measure should be included for which
4	program.
5	So, we are working on how to
6	implement it. Depends on the our
7	consultation with our leadership about which
8	program it's supposed to be. But I can see
9	that this measure could be for both programs.
10	Did I address your question about
11	implementation?
12	CO-CHAIR KAPLAN: I remain
13	confused. So if it's readmission to the
14	hospital following a vascular procedure that
15	could be done in either the outpatient or the
16	inpatient setting what we're looking at is the
17	readmission within 30 days of the procedure in
18	whatever setting it's done, is that correct?
19	DR. MCNAMARA: Yes. This is Bob
20	McNamara. I can address that. I appreciate
21	Lein giving us credit. I think we gave her a
22	headache with this decision.

	Page 114
1	We decided that really to be
2	clinically coherent and to make sense for the
3	clinical community to include both inpatients
4	and outpatients many of the procedures are
5	done as an outpatient procedure more based on
6	the hospital characteristics or provider
7	convenience or facilities rather than on
8	patients.
9	And to try to say, okay, we're
10	just going to do the inpatients then people
11	could come out of the measure just by changing
12	the setting, even if it's the same procedure.
13	So that was the logic behind it all.
14	To talk about that is to come into
15	the cohort you can have an outpatient
16	procedure at a hospital facility. But the
17	readmission has to be to the hospital. So
18	it's not another outpatient procedure for the
19	outcome.
20	CO-CHAIR KAPLAN: And for the
21	lumpers and splitters among us, so vascular
22	procedures seem like a big lumping category.

	Page 115
1	And to the extent that if you find
2	as these things roll out that some of these
3	vascular procedures look and behave different
4	from other vascular procedures is there a plan
5	when you're thinking about use in trying to
6	categorize smaller clumps?
7	DR. MCNAMARA: Well, I think
8	that's always a question of, as you said,
9	lumping and splitting, of how do you want to
10	do it.
11	The MedPAC had lumped them
12	together and I think that we thought that many
13	of the service lines within the hospitals, how
14	it's set up is such that one entity could
15	cover them all.
16	Certainly in the future some
17	people could pull out different ones but I
18	think the way practice is currently that it
19	made the most sense for us to include them
20	all.
21	CO-CHAIR KAPLAN: Thank you.
22	DR. ROBERTS: Is there a

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	Page 116
1	difference in risk for those vascular
2	procedures in inpatient versus outpatient?
3	DR. MCNAMARA: Yes, in a word.
4	There's differences across which procedures
5	you have. But the feeling was that it's not
6	necessarily, it was, you know, the patient
7	picking the patient's work, that lower risk
8	that would be done as an outpatient, not
9	necessarily that the facility as an outpatient
10	was what was causing them lower risk. So for
11	the patient level they were coming in as a
12	lower risk to be done as an outpatient.
13	But if a hospital has to choose
14	whether they do it as an outpatient or as an
15	inpatient I wanted to include them all because
16	it's really based upon the patient, not upon
17	the facility or location.
18	CO-CHAIR KAPLAN: Larry?
19	DR. GLANCE: Does your risk
20	adjustment model include an indicator for
21	whether the procedure was performed as an
22	inpatient versus outpatient?

	Page 117
1	DR. MCNAMARA: No. The way that
2	the cohort is developed was based upon coding.
3	And the inpatient codes are ICD-9 and the
4	outpatient codes are CPT. So, you could
5	develop it, you know, you could identify that
6	from there.
7	But again, basically because of
8	feeling that patients an individual patient
9	will get their procedure, or could get a
10	procedure as an inpatient and outpatient based
11	upon a facility rather than based upon patient
12	characteristics, it wouldn't be appropriate to
13	adjust inpatient versus outpatient. You're
14	trying to adjust it based upon the procedure
15	being done.
16	CO-CHAIR KAPLAN: Larry, did that
17	answer your question?
18	DR. GLANCE: Just a follow-up
19	question. Did you look at whether or not
20	there was a tendency for patients who were
21	procedures that were performed as outpatient
22	procedures, for the same procedures to be

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	Page 118
1	readmitted more often if they were performed
2	as an outpatient versus an inpatient?
3	In other words, the idea being
4	that the inpatient procedures, they're already
5	admitted to the hospital, whereas for the
6	outpatient procedures you're sort of in a
7	way you would think that those patients are
8	slightly more likely to be readmitted because
9	of being sent home more quickly.
10	DR. CURTIS: So this is Jeptha
11	Curtis. I can comment a little bit on that
12	specifically.
13	So, the question is whether or not
14	there's a downside to outpatient procedures.
15	And when we say outpatient procedure we're not
16	really talking about necessarily patients who
17	are going home the same day. Oftentimes we're
18	talking about procedures that are being
19	performed on an observation stay basis as
20	opposed to an outpatient stay. And that's
21	pretty much the major rationalization for
22	including them.

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	Page 119
1	So, most of the patients who are
2	outpatient or observation stay have the exact
3	same hospital utilization. They're in
4	overnight and they go home the next day. And
5	this is almost all endovascular procedures
6	being performed on the neck or in the legs.
7	And that's, you know, everything
8	is the same except for whether or not the
9	hospital administrator characterizes it as an
10	inpatient or an observation stay basis. So
11	there's really no other information that comes
12	with that.
13	The lower risk of readmission
14	associated with that observation stay
15	population really is driven by the fact that
16	there are low-risk populations no matter what
17	the designation is. And that's why we adjust
18	for the procedure, not for the setting.
19	CO-CHAIR KAPLAN: Thank you. Any
20	other comments?
21	DR. MCNAMARA: I'm sorry, it's Bob
22	McNamara. Just to add on that that there's

	Page 120
1	not a 1 to 1 correlation between the CPT codes
2	and the ICD-9 codes. You can't exactly say.
3	So certainly within there's many more CPT
4	codes, but they can be adjusted to different
5	ICD-9s. Or correlated.
6	CO-CHAIR KAPLAN: Thank you.
7	Paula?
8	MS. MINTON-FOLTZ: Can you tell me
9	if this excludes same-day transfers? So if I
10	came in as an outpatient procedure and was
11	immediately admitted afterwards would that
12	count as a readmission?
13	DR. MCNAMARA: I believe it would
14	not. We included that it was a 24-hour
15	difference in the level or the date of the
16	procedure and the admission.
17	And I think most of those within
18	Medicare rules would be listed as an inpatient
19	procedure, even if as Jeptha said, many of
20	these are done in the same area. It could be
21	done in the same operating suite, whether it's
22	an inpatient or an outpatient officially.

	Page 121
1	And then even if they were intake
2	to the surgery as an outpatient, if they
3	decided to change that as an inpatient from
4	the back end Medicare it would look like an
5	inpatient, not as a readmission.
6	CO-CHAIR KAPLAN: Thank you.
7	Karen and then Karen.
8	DR. MCNAMARA: That was Bob
9	McNamara again.
10	MS. PACE: Right, this is Karen
11	Pace. I just wanted to make a comment that we
12	really wouldn't want to include a risk factor
13	related to where the procedure took place
14	because that may be one of the things that
15	might be a difference in the care provided.
16	So it may be something that's
17	useful for drilldown and quality improvement
18	when you're looking at your data and what
19	patients are being readmitted. But it
20	generally wouldn't be something that would be
21	considered for a risk factor.
22	CO-CHAIR KAPLAN: Karen?

	Page 122
1	DR. JOYNT: I just want to say
2	something kind of similar which is just to
3	commend you for going to what I'm sure was a
4	lot of trouble to put the outpatient things
5	in. I think that is hugely important,
6	especially as care sort of shifts place to
7	really think about quality spanning across
8	different settings.
9	I just have more of a technical
10	question maybe for NQF people which is with
11	this measure, compared to others who really
12	don't have a clue for how it's going to work
13	in terms of the outliers. Is that something
14	that we are expected to know as we think about
15	whether or not we feel that the measure is
16	appropriate?
17	Or do we just sort of say it
18	doesn't matter if it identifies 4 percent or
19	even 25 percent as outliers, that's separate
20	from the measure itself? Any guidance would
21	be helpful.
22	MS. PACE: So, I guess we could

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	Page 123
1	look at the the information that they
2	provided was just kind of the distribution of
3	the scores. And we could look at that.
4	But I don't think that has to be
5	your defining decision. You may want to not
6	say high on usability and use because that's
7	a question in your mind, but the real question
8	is whether this has the potential at this
9	point to be useful for improvement and
10	accountability.
11	And you know, one of the things
12	when it comes back for endorsement maintenance
13	will be to have some real data on how that has
14	played out. So I don't think it's an ultimate
15	defining decision for use and usability.
16	CO-CHAIR KAPLAN: Yes. So the way
17	I understand this process. Correct me if I'm
18	wrong, really quickly. If we endorse this for
19	use then it goes out and they get the
20	information, Karen, that you would be looking
21	for within 3 years or they don't. And when it
22	comes back for re-approval then we reconsider

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	Page 124
1	whether or not the information is suitable for
2	re-endorsing this measure? Does that help?
3	DR. JOYNT: One more question.
4	With all these measures that are again sort of
5	a longer time frame and the information is not
6	fed back to the hospital quite as quickly as
7	might be optimal for quality improvement is
8	that again something we should consider in
9	approving a metric, or something that goes to
10	ways that we hope that all these measures get
11	used better in the future?
12	Because I think that's a
13	limitation that cuts across a lot of these.
14	It's nothing to do with the statistical power
15	of the model, or the way that it's set up, or
16	the way that you've chosen procedures. It's
17	just I think it's a real problem if we're
18	supposed to also think about the usability for
19	improvement. So how should we think about
20	that?
21	CO-CHAIR KAPLAN: That's also come
22	up before. I'll let you go ahead.

	Page 125
1	DR. BURSTIN: This has come up in
2	multiple discussions for us, particularly
3	around the readmission measures and the lag
4	time to get the amount of time back.
5	Again, I think it's something you
6	could factor into usability as you're voting.
7	I also know it's something Lein and others
8	from CMS have pointed out as something you're
9	actively working on, trying to maybe Lein
10	wants to respond.
11	But I know there have been active
12	efforts to see if there are more ways to get
13	information back to hospitals more quickly.
14	Lein?
15	DR. HAN: This is Lein Han from
16	CMS. We got this feedback all the time from
17	the hospitals. And it's understandable that
18	they do need most data for quality
19	improvement.
20	So, what we're working on is that
21	we're not providing the risk-adjusted rate
22	quarterly, but we would like to see if we can

	Page 126
1	just get the data to them. But the raw data,
2	not really the calculated data. So when they
3	have the raw data at least they can look at
4	the cases.
5	So, and we plan to do this
6	quarterly, hopefully that we can get to the
7	hospital quarterly this type of data. But raw
8	data like we provided in hospital-specific
9	report, those cases. Thanks.
10	CO-CHAIR KAPLAN: Thank you.
11	Karen, one of the issues that's come up before
12	is that in that backwards look for three years
13	time, for example, the tensions between
14	getting a precise estimate and so you get more
15	cases means that you lose on the other end in
16	terms of usability for quality improvement
17	issues. So there are some tensions and
18	tradeoffs in these different kinds of calls.
19	DR. BRIGGS: So in the reporting
20	is this going to be reported out by the
21	anatomical buckets? Or is this going to be
22	vascular readmissions altogether?

	Page 127
1	DR. MCNAMARA: As the measure
2	specifies right now there will be an overall
3	vascular readmission.
4	Whether in the future it can be
5	done on a procedure level for some of the
6	high-volume procedures or in different buckets
7	can be done based upon how it's set up. But
8	the measure was developed as an overall.
9	DR. CURTIS: Just to follow up on
10	this. Jeptha Curtis. I think the way that
11	you could use it, I think you'd report out all
12	vascular readmission rate and that's useful
13	for public reporting.
14	To the hospitals we could try and
15	create buckets that make it more usable for
16	them for actually driving quality improvement
17	processes so they can know where they are
18	maybe not in a risk-adjusted fashion but at
19	least what's driving their hospital-specific
20	readmission rates.
21	CO-CHAIR KAPLAN: Okay, we're
22	coming up on time. I don't want to cut this

	Page 128
1	short. Any other comments or questions?
2	We're ready to vote on usability.
3	MS. SHAHAB: Voting for usability
4	and use. One high, two moderate, three low,
5	four insufficient information and the time
6	begins now.
7	We have all the votes for
8	usability and use. One high, eleven moderate,
9	four low, four insufficient information.
10	CO-CHAIR KAPLAN: Thank you and I
11	think we're ready to vote on endorsement. Any
12	comments?
13	Okay, ready to vote.
14	MS. SHAHAB: voting for overall
15	suitability for endorsement. One yes, two no.
16	Time begins now. Just one more vote. Can you
17	please just press your votes one more time?
18	We have all the votes for overall
19	suitability for endorsement for measure 2513
20	Hospital 30-day All-cause Risk-standardized
21	Readmission Rate Following Vascular
22	Procedures. The votes are 14 yes, 6 no.

	Page 129
1	CO-CHAIR KAPLAN: Thank you very
2	much to the developers for coming and for CMS
3	coming as well.
4	And we finished within three
5	minutes which is measurement error in my view
6	on time. So, excellent. We we have a break
7	until 10:15. Thank you.
8	(Whereupon, the foregoing matter
9	went off the record at 10:03 a.m. and went
10	back on the record at 10:13 a.m.)
11	CO-CHAIR KAPLAN: Can we ask our
12	developers to briefly introduce yourselves?
13	And make sure when you're commenting that you
14	state your name and briefly give us a two-
15	minute brief discussion of the measure.
16	DR. NAKAMURA: Thank you. My name
17	is Mari Nakamura. I'm a pediatric infectious
18	diseases doctor and health services researcher
19	at Boston Children's Hospital.
20	DR. ZASLAVSKY: I'm Alan
21	Zaslavsky. I'm a statistician at Harvard
22	Medical School.

	Page 130
1	DR. NAKAMURA: And as you heard
2	joining us on the phone is our principal
3	investigator for our center Mark Schuster
4	who's joining us from Vancouver today.
5	Measuring and reducing
6	readmissions has become a widespread focus in
7	pediatrics, but to date no readmission
8	measures developed specifically for use in
9	children and adolescents have been publicly
10	available.
11	We were therefore assigned to
12	develop readmission measures by CMS and AHRQ
13	as part of their pediatric quality measures
14	program for which we serve as the center of
15	excellence.
16	Hospital readmissions within 30
17	days occur for 2 to 6 percent of children.
18	These rates are certainly lower than the rates
19	of about 20 percent that we often hear for
20	Medicare beneficiaries over age 65, but
21	overlap with rates for adults under age 65.
22	As a point of comparison pediatric

	Page 131
1	30-day readmission rates are equivalent to
2	pediatric inpatient adverse drug event rates.
3	Hospitals, payers and other
4	stakeholders are actually already actively
5	working to reduce pediatric readmissions even
6	in the absence of a publicly available
7	measure.
8	Our all-condition measure
9	evaluates readmissions following an index
10	hospitalization for almost any condition.
11	We were encouraged by CMS to
12	develop an all-condition measure to correspond
13	with the adult measure that they've now rolled
14	out. And in addition, our national
15	stakeholder panel supported an all-condition
16	measure because it includes the broadest range
17	of children and hospitals.
18	Furthermore, we found that very
19	few specific pediatric conditions are common
20	enough to serve as a focus of a readmission
21	measure.
22	We've prioritized harmonizing our

Page 132
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measure with the NQF-endorsed adult
readmission measures while still making it
appropriate for pediatric use.
One important way in which our
measure corresponds with NQF-endorsed adult
measures is that we choose to focus on
evaluating unplanned readmissions. Based on
our own research as well as other studies we
don't think that the preventability of
readmissions can be assessed using billing
codes.
The main data set we used to
develop and test the measure consisted of
Medicaid claims for about 400,000
hospitalizations at 2,000 hospitals in 26
states.
We also used AHRQ HCUP all-payer
data from 2 states and NACHRI case mix data
from 72 children's hospitals.
We developed a case mix adjustment
model for the measure that includes patient
age, gender and chronic conditions on the

	Page 133
1	index hospitalization and we found that the
2	model performed similarly to those used in
3	other readmission measures with regard to
4	discrimination and calibration.
5	A challenge not just for
6	calculating pediatric readmission rates but
7	for all pediatric quality measurement is small
8	sample sizes at some hospitals with resulting
9	low reliability of measure scores.
10	However, because pediatric
11	patients are not distributed across as many
12	hospitals as adult patients we found that the
13	majority of pediatric hospitalizations occur
14	at higher-volume hospitals whose readmission
15	rates have good reliability.
16	Because the measure uses claims
17	data that are already collected for other
18	purposes we anticipate that implementing it
19	will be highly feasible.
20	We believe that the measure fills
21	an important need for publicly available
22	readmission measures and think that it could

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	Page 134
1	serve as a valuable tool to assess health
2	system quality and motivate improvements in
3	pediatric care delivery. Thank you.
4	CO-CHAIR KAPLAN: Thank you very
5	much. Kathy?
6	DR. AUGER: So, to speak to
7	evidence. Certainly there isn't as much
8	evidence around pediatric readmission as there
9	is in the adult world.
10	However, certainly it meets the
11	overall construct that this could be an
12	important measure. So I think it's high in
13	that sense.
14	CO-CHAIR KAPLAN: Karen, do you
15	have anything to add? Are we ready to vote
16	evidence? Any other discussion? Go.
17	MS. SHAHAB: Voting for 1(a)
18	evidence. One is yes, two is no. And time
19	begins now. We need one more vote, please.
20	We have all the votes for
21	evidence. Twenty-one yes, one no.
22	CO-CHAIR KAPLAN: Thank you.

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	Page 135
1	Performance gap. Kathy?
2	DR. AUGER: So, as the developers
3	note the prevalence of pediatric readmission
4	rate is between 2 and 6 percent.
5	What I was just looking for and
6	couldn't find is what the range in
7	interquartile ranges for the risk-standardized
8	rate. Do you guys have that available?
9	DR. NAKAMURA: This is Mari
10	Nakamura again. To give you a sense for an
11	all-condition measure based on the variance
12	component of the hospital random effect in our
13	mix model a hospital that's two standard
14	deviations below the mean would have a
15	readmission rate of 2.4 percent whereas one
16	that's two standard deviations above would
17	have a readmission rate of 10.3 percent. So
18	about 4 times greater.
19	CO-CHAIR KAPLAN: Karen?
20	DR. JOYNT: We're actually looking
21	for this information together. And I think
22	that there's actually impressive performance

Page 136 1 gap when looked at that way. The graph -- I didn't find a 2 distribution on this. The distribution in the 3 Berry paper from last year suggests that most 4 hospitals actually don't fall that far outside 5 the mean. So do you know the 25th and 75th 6 7 percentiles? DR. NAKAMURA: I don't have that 8 9 with me, no. 10 DR. JOYNT: In that one it looked 11 like there were really very few that were past about between 5 and 7. But that may be 12 13 because it was the NACHRI hospitals. DR. NAKAMURA: This is Mari 14 Nakamura again. So, it's a good point that in 15 16 that study it was a quite homogenous set of 17 hospitals. And so you might expect that there wouldn't be as much difference among them in 18 terms of the range of readmission rates. 19 20 CO-CHAIR KAPLAN: I might point out that the magnitude of the difference is 21 about in the range of other measures that 22

	Page 137
1	we've looked at before. In fact, it seems to
2	be somewhat broader than some of them that we
3	looked at.
4	Other comments? Ready to vote
5	performance gap?
6	MS. SHAHAB: Voting for 1(b)
7	performance gap. One is high, two moderate,
8	three low, four insufficient and the time
9	begins now. One more vote, please.
10	We have all the votes for 1(b)
11	performance gap. One high, twenty moderate,
12	one low, zero insufficient.
13	CO-CHAIR KAPLAN: Great.
14	Priority?
15	DR. AUGER: So, the developers
16	mentioned cost of readmissions at six months
17	which is certainly a longer window than the
18	measure in front of us which was \$136 million.
19	So, of course pediatric costs are
20	not anything what adult costs are. Having
21	said that, they do also present data on
22	disparities that exist in pediatric

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	Page 138
1	readmission, including race/ethnicities
2	disparities and payer disparities. So that
3	goes certainly to priority.
4	And then also of course as they
5	mentioned this would be the first pediatric
6	metric. So that again speaks to priority.
7	CO-CHAIR KAPLAN: Karen? Other
8	comments? Ready to vote?
9	MS. SHAHAB: 1(c) high priority.
10	One high, two moderate, three low, four
11	insufficient. Time begins now. One more.
12	We have all the votes for 1(c)
13	high priority. Seven high, thirteen moderate,
14	two low, zero insufficient.
15	CO-CHAIR KAPLAN: Thank you.
16	Scientific acceptability. We'll talk about
17	reliability first. Kathy?
18	DR. AUGER: Sure. So, as the
19	measure development experts had mentioned
20	previously they used ICC to assess
21	reliability. And it's very much dependent on
22	the volume of cases seen at hospitals.

Page 139 1 So, with the reliability of 0.5 there were only 607 out of the 2,011 hospitals 2 3 that had an ICC greater than that but that did a count for 88 percent of the index 4 hospitalizations. So the majority of the 5 hospitalizations are at higher-volume 6 hospitals which had the higher reliability 7 with the low-volume hospitals are the issue. 8 9 CO-CHAIR KAPLAN: Karen? Other 10 comments? Ready to vote reliability? 11 MS. SHAHAB: Voting for 2(a) reliability. One is high, two moderate, three 12 13 low, four insufficient and the time begins 14 now. One more. We have all the votes for 2(a)15 16 reliability. Three high, seventeen moderate, 17 two low and zero insufficient. 18 CO-CHAIR KAPLAN: Thank you. Validity. Kathy? 19 20 DR. AUGER: So, this is an 21 examination of unplanned hospitalization or 22 readmissions. So the developers have a

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	Page 140
1	somewhat novel way of determining planned or
2	unplanned which is worth mentioning.
3	The way that they did that was
4	through expert opinion panels of which codes
5	could be consistent with a planned procedure.
6	So they also went through some validation of
7	that algorithm in and of itself using chart
8	review at Boston Children's which seemed
9	reasonable.
10	Then in terms of the risk
11	adjustment validity they use the number age
12	and the number of chronic conditions as well
13	as gender.
14	There was some concern in the
15	public reporting comments that they hadn't
16	used primary diagnosis so that might be
17	something worth just asking about.
18	And then in terms of how the model
19	performed the C statistic was 0.69. In terms
20	of calibration there was good observed-to-
21	expected graphs.
22	And then the other question I had

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	Page 141
1	was again about how many hospitals or
2	outliers, whether or not they have done that
3	assessment.
4	And finally, other threats to
5	validity would just be in terms of missing
6	data because of the MAX chart. The MAX data
7	system in and of itself has a lot of issues
8	with handling admission data.
9	CO-CHAIR KAPLAN: Karen, I'm going
10	to ask you to hold on a second and let the
11	measure developers respond.
12	DR. NAKAMURA: Thank you.
13	Regarding the question of
14	CO-CHAIR KAPLAN: State your name,
15	please.
16	DR. NAKAMURA: Sorry. Mari
17	Nakamura. Regarding the question of including
18	the reason for admission, the primary
19	diagnosis, this was something that we thought
20	a great deal about and had done some
21	exploratory analysis of and ultimately
22	concluded didn't make sense to include.

	Page 142
1	One reason is that we found that
2	actually performance in a given hospital
3	tended to track across different types of
4	diagnoses so that it does make sense to be
5	able to aggregate all of the patients into a
6	single measure.
7	We also found that there's a
8	challenge in using pediatric diagnosis codes
9	with a good grouping system. We even
10	experimented with trying to devise one of our
11	own.
12	Because patients don't have just a
13	sort of few common diagnoses but really in
14	pediatrics diagnoses are quite variable we
15	found that for any given category that for
16	some hospitals there were problems with cell
17	sizes.
18	I don't know if Alan has anything
19	further he might want to say? No? Okay.
20	Regarding the question of outliers
21	what we had provided in our submission was one
22	way of looking at outliers which is to use the

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	Page 143
1	method that's currently used for the adult
2	measure pay-for-performance program.
3	And using those methods we found
4	that for the all-condition readmission measure
5	that about 47 percent of hospitals have a
6	higher than expected readmission rate. So
7	their predicted readmissions exceed their
8	expected.
9	And that the median readmission
10	rate for those hospitals that were above 1 in
11	their ratio was 1.15 suggesting that they have
12	about 15 percent in excess in terms of the
13	median readmissions.
14	We haven't looked at the question
15	in terms of outliers using confidence
16	intervals, but recognize that there are
17	different ways that one could apply the
18	measure and choose to identify outliers.
19	And then I think your last
20	question was about missing data in MAX. So we
21	definitely acknowledge that MAX is a very
22	messy data set as we learned once we delved

	Page 144
1	into it.
2	For those of you who may not be as
3	familiar it's assembled by collecting Medicaid
4	claims from all of the 50 states and the
5	District of Columbia and then trying to make
6	them into a uniform data set for researchers
7	to use.
8	And as you might imagine that
9	process is a difficult one. It seems to be
10	definitely improving over time. But that is
11	one reason, for example, that in our MAX data
12	set we felt that not all states had good
13	enough data quality for key variables such as
14	hospital identifiers to be able to include
15	them.
16	So all that said while our test
17	data set was the MAX data and we were actually
18	pleasantly surprised that the percentage of
19	records that had to be dropped based on data
20	quality or completeness issues was actually 10
21	percent.
22	That doesn't necessarily mean that
	Page 145
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1	the real data sets that would be used, meaning
2	the actual claims data available to, for
3	example, state Medicaid agencies would have as
4	many issues with missing data as the MAX data.
5	CO-CHAIR KAPLAN: Thank you.
6	We're going to go Karen and then Frank.
7	DR. JOYNT: Yes, just had a few
8	additional questions to the ones that were
9	brought up before.
10	I think one of the threats to
11	validity is just that there aren't other
12	pediatric measures with which to compare this
13	one. And so it's a little bit difficult to
14	know exactly what we're measuring.
15	Certainly hospitals can differ on
16	things like socioeconomic status and access to
17	care for kids. And that may be the difference
18	that we're seeing driving this. It's hard
19	without having anything else that we would
20	sort of consider to be "quality" to compare
21	this to to know exactly what we're measuring
22	here.

	Page 146
1	I think that's probably the case
2	with most of the readmission metrics to some
3	degree but it's particularly problematic if we
4	don't have comparisons. This is not the
5	developer's fault and kudos to you for trying
6	to develop a quality metric in what is often
7	a very data-free zone. But I think it is an
8	important threat to validity.
9	The two other well, I guess the
10	one other question is really if you could just
11	explain this a little bit, how this model is
12	similar or different to the ones that we're
13	used to hearing for the adult metrics. Just
14	so that we are clear on whether or not this is
15	the same method as is being used for the other
16	measures or if it's different and in what ways
17	it differs beyond the exclusion of procedures.
18	Or, sorry, planned admissions.
19	DR. NAKAMURA: Thank you. This is
20	Mari Nakamura again. Karen is absolutely
21	right that a challenge we faced in trying to
22	assess the validity of our measures is the

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	Page 147
1	fact that there are not other pediatric
2	inpatient measures we could use, or widely
3	available data sources for which we would be
4	able to even evaluate such measures.
5	We agree that this is a really
6	important question and one that we think will
7	need to be evaluated as more measures are
8	developed. We felt like it was good to start
9	somewhere and acknowledged that this is a
10	limitation currently in our field in pediatric
11	measurement.
12	Regarding the question of how our
13	measure compares to the adult measures in
14	terms of our statistical approach. Overall
15	the approaches are really similar in that we
16	use hierarchical modeling.
17	We do as a result have the
18	shrinkage effect that I've heard discussed
19	quite a bit here. And we've talked about both
20	the advantages of that.
21	It is relevant to pediatrics
22	because of course we do have many small-volume

	Page 148
1	hospitals in terms of pediatric volume.
2	One difference with the all-
3	condition measure is that the all-condition
4	measure for the adult Yale measure uses five
5	different service line models and then
6	combines the outputs of that to end up with a
7	single readmission rate.
8	We considered such an approach but
9	our worry was that at many hospitals given
10	that pediatric volumes overall are low that we
11	would then have trouble with the sample sizes
12	for splitting our sample among different
13	models.
14	Another difference which Alan may
15	want to speak a little bit more to in terms of
16	the statistical implications is that we use
17	direct rather than indirect standardization.
18	Meaning that in our approach of
19	standardizing we hypothesize what the rate
20	would be at a given hospital assuming that the
21	entire cohort of reference data set was cared
22	for at that hospital. In the indirect method

	Page 149
1	instead it's this approach of using predicted
2	to expected readmissions.
3	But the end result again is that
4	it tends to pull the small-volume hospitals
5	closer to the mean. So in that way the
6	outputs are similar.
7	DR. ZASLAVSKY: Alan Zaslavsky. I
8	would just add that while the direct and
9	indirect standardization look on the face of
10	it pretty different the underlying models
11	actually work out to be pretty similar.
12	The direct standardization we're
13	doing uses the logistic model. Indirect uses
14	usually a ratio of observed to expected which
15	implies a multiplicative or log linear model.
16	But in the range we're talking about the two
17	models are pretty close to each other. So,
18	that doesn't make a big difference.
19	And the shrinkage effects as Mari
20	said are handled in pretty similar ways. The
21	sufficient statistic for the performance of a
22	particular hospital is essentially the total

	Page 150
1	number of readmissions. And that's the same
2	in both models. So the two should give pretty
3	similar results.
4	We're another group. We set
5	things up a little differently in a way that
6	we found to be a little bit more direct, not
7	just because it's direct standardization but
8	because it's all done on one model. But I
9	think the two are similar enough that we
10	wouldn't have found anything terribly
11	different if we'd done it exactly the way Yale
12	did it.
13	CO-CHAIR KAPLAN: Thank you. Any
14	other comments? Frank, and then Leslie, and
15	then Tony.
16	DR. BRIGGS: So, two quick
17	questions. First just being a definition.
18	You said you excluded specialty hospitals. I
19	was wondering if that was the same cancer
20	hospitals that you see in the adult realm.
21	And then the other was hospitals,
22	if the readmission was in the discharge was

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	Page 151
1	readmitted to an area out of the state because
2	of limitations on the data set. Although you
3	might have some data I was wondering how often
4	that happened, especially for border states
5	and rural care and things of that nature.
6	DR. NAKAMURA: So for the question
7	about specialty hospitals, in pediatrics that
8	designation tends to capture, for example,
9	hospitals that do deal with specific
10	conditions. Cancer hospitals are one.
11	Another would orthopedic hospitals.
12	The other sort of major category
13	is more and another category, excuse me, is
14	the Shriner's hospitals that deal with burns
15	and trauma.
16	And then another sort of group of
17	hospitals that we excluded is hospitals that
18	don't provide acute care, that are more long-
19	term care such as for rehabilitation.
20	For the question about what
21	percentage of readmissions are to an out-of-
22	state hospital, so the good thing is that

Page 152 1 because Medicaid claims go back to the state of residence of the patient there's actually 2 3 complete data for the patients who are in a given state about where they were readmitted. 4 Where there is a challenge is for 5 the states that weren't in our data set. 6 Because for those we have probably a minority 7 of information. We would only know about he 8 9 claims that happen to come through the 10 particular states that were in our data set. 11 And so that's why we exclude the hospitals that are outside of the states that 12 13 we include in our data set. But for those that are in the data set we would be able to 14 tell if patients are admitted outside of their 15 16 home state. 17 CO-CHAIR KAPLAN: Thank you. Leslie? 18 So, I have a question. 19 MS. HALL: 20 Where you might have two hospitals in the same 21 community, one hospital is known as a brand for children's hospital. It's not a 22

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	Page 153
1	children's hospital, it's just where most
2	children go.
3	Both hospitals would be very high-
4	volume community hospitals. But one would
5	display as a low-volume hospital even though
6	it wasn't in this case. Does this create any
7	challenges where we might have the same sort
8	of problem we've mentioned over and over again
9	about low-volume hospital size and low-volume
10	overall when we just simply have a patient mix
11	that's very different in an otherwise high-
12	volume hospital?
13	DR. NAKAMURA: This is Mari
14	Nakamura. I want to make sure I understand
15	your question.
16	So you're wondering about the
17	issue of community hospital that has done
18	pediatric patients but overall not a very high
19	volume of pediatric patients compared to
20	another hospital?
21	MS. HALL: That's correct.
22	DR. NAKAMURA: So, this is

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	Page 154
1	certainly a challenge about trying to measure
2	pediatric readmission rates or any sort of
3	hospital-based quality measure. You're right
4	that there are big differences in the types of
5	hospitals that care for children.
6	What we've found is that children
7	tend to be really concentrated at relatively
8	small numbers of hospitals. So they tend to
9	be high-volume, they tend to provide a full
10	range of care. And then the community
11	hospitals tend to provide care for relatively
12	small numbers of children.
13	To give you a sense, children's
14	hospitals make up about 5 percent of all of
15	the 4,000 hospitals in the country but
16	actually care for about one-third of pediatric
17	inpatients.
18	And so I think where this is a
19	challenge is that as we've alluded to for
20	these community hospitals on the one hand we
21	didn't want to ignore them. But we also
22	acknowledge that the reliability of their

	Page 155
1	readmission rates is limited because of the
2	low volume.
3	And so one way that we think these
4	hospitals could still be included is of course
5	to be very responsible about explaining the
6	limits and what we actually know about them.
7	And also perhaps to compare like hospitals
8	with like.
9	CO-CHAIR KAPLAN: Thank you.
10	Tony?
11	DR. GRIGONIS: Yes, I just have a
12	quick question about your databases from you
13	said 26 states I believe. Were they
14	consistent in terms of the state populations?
15	That's the first question.
16	The second follow-up would be did
17	you see any differences in states that have
18	higher populations.
19	DR. NAKAMURA: So to answer
20	this is Mari Nakamura again to answer the
21	question about population, the states actually
22	varied quite a bit in population. Because we

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	Page 156
1	found that the states with the best data
2	quality fortunately were geographically
3	distributed. And so some of them were lower
4	population states and other high.
5	We didn't actually do an analysis
6	to evaluate how the volume of the state
7	related to readmission rates. That's
8	something we certainly would be able to do.
9	CO-CHAIR KAPLAN: Thank you very
10	much. Other comments?
11	I would remind the group that as
12	with the previous measure on all-cause
13	readmissions following vascular procedures
14	this would be a new measure. So if endorsed
15	the 3-year period would ask that the
16	developers generate some of the data that have
17	been raised and issues of concern and so on.
18	So if approved it would be a new
19	measure that we would ask the developers to
20	consider some of these response to some of
21	these kinds of issues.
22	Are we ready to vote validity?

	Page 157
1	MS. SHAHAB: Voting for 2(b)
2	validity. One is high, two moderate, three
3	low, four insufficient. Time begins now.
4	We have all the votes for 2(b)
5	validity. Zero voted high, nineteen voted
6	moderate, three low and zero insufficient.
7	CO-CHAIR KAPLAN: Thank you.
8	Feasibility. Kathy?
9	DR. AUGER: Certainly these are
10	claims data so potentially feasible.
11	I think it is worth mentioning, as
12	Mari already did, that what's in Medicaid
13	claims does vary from state to state. So
14	that's a little bit of an issue.
15	Having said that they did test the
16	measure on the New York database and were able
17	to see a little bit of model fitting issues
18	with rare values. But then they were able to
19	offer some troubleshooting to address any
20	issues of feasibility. So I think that's a
21	strength.
22	CO-CHAIR KAPLAN: Karen?

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	Page 158
1	DR. JOYNT: Yes, I think
2	feasibility is a real concern here because
3	kids, unlike adults over 65, are covered by a
4	whole bunch of different insurance plans.
5	So yes, there's a chunk in
6	Medicaid but they certainly don't represent a
7	randomly selected group of kids. And so
8	understanding how this measure might act in
9	all-payer claims databases versus a Medicaid
10	database versus Aetna or something like that
11	I think is something that would really need to
12	be thought through as this is rolled out.
13	Are we trying to build a Medicaid
14	quality metric? Are we trying to build a
15	pediatric quality metric? And I think those
16	two things are probably different. And so I
17	think the model and the measure would probably
18	have to take that into account given the
19	limitations of the pediatric data and the ways
20	in which the populations that underlie those
21	data might actually differ quite a bit.
22	CO-CHAIR KAPLAN: Other comments

	Page 159
1	or thoughts?
2	DR. NAKAMURA: Thank you. This is
3	Mari Nakamura again.
4	The issue about the variation in
5	Medicaid claims from state to state is
6	certainly a true one. We anticipate that one
7	of the likely uses for the measure will be by
8	Medicaid programs. We know that's something
9	that CMS is interested in.
10	And at least initially it seems
11	that what would be most feasible is to
12	evaluate readmission rates within a state
13	rather than trying to compare across states.
14	That said it may be that as more
15	measures come out for pediatrics that this
16	will perhaps drive the creation of a national
17	data set that's available in a more timely way
18	than the MAX data set.
19	We also note that there are
20	several states that are developing or already
21	have all-payer claims data sets that could be
22	useful for this measure.

	Page 160
1	Again, that wouldn't necessarily
2	allow national comparisons but at least
3	within-state comparisons.
4	One thing that we tried to do with
5	our measure and have provided the SAS program
6	for is we've provided a way for a given state
7	Medicaid program to be able to calculate
8	nationally comparable rates using our MAX data
9	set as a reference data set.
10	And so this is something that if a
11	state Medicaid program really wanted to have
12	a sense of how they compared to another state
13	that they would have the option to use.
14	Regarding Karen's point about the
15	fact that there are different insurers for
16	children and that unlike the adult over-65
17	program there's not a Medicare for children.
18	This is very true.
19	Medicaid covers about one-third of
20	hospitalized children. So it is a sizeable
21	portion. But she's absolutely right that
22	those children are not necessarily directly

	Page 161
1	comparable to children covered by other
2	insurance plans.
3	We've looked at insurance
4	disparities and found in fact that Medicaid-
5	insured children do have a higher risk of
6	readmission.
7	We know that there has been a lot
8	of discussion about including such sort of
9	sociodemographic factors. And if there was
10	interest in including something like insurance
11	status for pediatrics we would be very
12	interested in discussing that further.
13	CO-CHAIR KAPLAN: Thank you. I'm
14	going to we're shading into usability here
15	so I'm going to ask Tom and then Larry.
16	DR. SMITH: Well, that was my
17	question. Is this a feasibility issue or a
18	usability issue? Are all your data from MAX
19	data? And are you putting this out there as
20	a Medicaid performance measure or a children's
21	performance measure?
22	DR. NAKAMURA: This is Mari

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	Page 162
1	Nakamura again. As a pediatric performance
2	measure but recognizing that one probably
3	common use will be as a Medicaid performance
4	measure. We were aiming to be as inclusive as
5	we could.
6	An example of another data set
7	that could be used with our measure would be
8	an all-payer claims data set at a state level.
9	And so as part of our measure testing we did
10	work with the State of New York as Kathy
11	mentioned to test the measures on their all-
12	payer data set as well.
13	CO-CHAIR KAPLAN: Again, this is a
14	usability issue. We're still on feasibility.
15	DR. SMITH: Well, I was going to
16	say I now agree with Karen that there's a
17	significant feasibility issue here based upon
18	your response.
19	CO-CHAIR KAPLAN: So feasibility
20	for application across all I'm concerned
21	about your issue. Let me have you guys re-
22	frame that as what is the actual nature of the

	Page 163
1	concern about feasibility?
2	DR. SMITH: I'll defer to Karen.
3	DR. JOYNT: Well, I just I
4	think it depends what you're trying to define
5	it as. As a Medicaid measure it is feasible
6	for states to use because it's based on claims
7	data.
8	CO-CHAIR KAPLAN: So is your
9	concern the target of inference here?
10	DR. JOYNT: As a pediatric measure
11	it's not super feasible because there are
12	entire swaths of the population that are in no
13	claims data set. Which makes it very
14	difficult to apply a claims-based metric to
15	them. So I think it depends a little bit on
16	the population that we're trying to assess
17	whether or not it is actually feasible to do.
18	If you had a kid in, I don't know,
19	Kansas I don't know if Kansas is an all-
20	payer claims data set covered by Aetna I'm
21	not sure that you actually could include that
22	child in a metric. But for Medicaid it's

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	Page 164
1	highly feasible because it's a claims base.
2	So it just depends on the group.
3	CO-CHAIR KAPLAN: Larry? Taroon?
4	MR. AMIN: It's an interesting
5	sort of characterization of the feasibility
6	question.
7	The way that we generally think
8	about it is so if a state had an all-payer
9	claims database, let's say California for
10	instance, and they wanted to be able to run
11	this measure for their pediatric patients in
12	their state, would they be able to do that?
13	I think that's generally the way that we would
14	look at it. So if they had the data can they
15	take these specifications and run them?
16	And so I'll just open that. And
17	if they can I think that makes it a very
18	feasible measure. I mean, if there's no if
19	a user doesn't even have the data that's not
20	necessarily a feasibility question for the
21	measure developer.
22	So, maybe I'll just with that

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	Page 165
1	framing maybe the measure developer can
2	respond.
3	DR. NAKAMURA: Yes, thank you.
4	This is Mari Nakamura.
5	So with that framing we certainly
6	think that the measure is highly feasible,
7	that it's actually quite easy to implement.
8	We have very detailed specifications and we've
9	also provided or can provide programs to
10	actually be able to do a lot of the data
11	preparation and running of the model.
12	So in our experience working with
13	New York and in working with the programs
14	ourselves we feel like if you have the data
15	that it's highly feasible.
16	CO-CHAIR KAPLAN: Thank you.
17	Larry?
18	DR. GLANCE: A quick comment
19	though. Was the model validated in all-payer
20	data? And if not, although it may be feasible
21	to use with all-payer data it probably would
22	not be a very appropriate use if it has not

	Page 166
1	been validated using all-payer data.
2	DR. NAKAMURA: It's true that our
3	primary data set for development was Medicaid
4	claims data. And so we've done the most
5	testing absolutely in that data set.
6	We did also, however, use a couple
7	of states, HCUP state inpatient database data
8	to test the measures as well. And to be able
9	to also in the case of New York to compare our
10	findings with their findings on their all-
11	payer claims data set.
12	CO-CHAIR KAPLAN: So now I would
13	like to ask NQF to comment on how we frame
14	this. Because the issue has become the target
15	of inference here.
16	And if we're trying to infer what
17	the all-cause readmission rates for pediatrics
18	are using the Medicaid database that's a
19	different issue I'm hearing that's a
20	different issue than if you're trying to
21	extrapolate to the entire pediatric
22	population.

	Page 167
1	MR. AMIN: Well, I'll answer but
2	with a question I guess. I mean, in some ways
3	when we look at I mean, not to draw
4	comparisons with other measures, but you know,
5	the Yale CMS measures that are the all-
6	cause hospital readmission measure, has been
7	developed using Medicare claims data, then
8	tested with California all-payer claims
9	database. So, and it's specified as 18 and
10	older. So the unit of inference could be any
11	group of people that are over that population.
12	In this case it seems very similar to me.
13	So I would say that I mean, I'm
14	not judging this but I think you should keep
15	that in mind as you make your decision about
16	the feasibility of this measure.
17	CO-CHAIR KAPLAN: Leslie?
18	MS. HALL: So, just a
19	clarification though. Didn't you already
20	state that the research showed that there
21	actually was a difference with the current
22	Medicaid covered population, and that you saw

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	Page 168
1	right away that there was a difference in that
2	insured population versus the other insured
3	population?
4	So therefore we're starting to
5	game with a difference in class of service I
6	guess based upon the insurance. And so do we
7	have mixed messages?
8	DR. NAKAMURA: Thank you. This is
9	Mari Nakamura. So yes, you're correct that
10	Medicaid-insured children have a higher
11	readmission risk than privately insured
12	children, for example.
13	This would be one argument for why
14	it might be important to eventually adjust for
15	insurance status as part of the case mix
16	adjustment.
17	At the same time, to go back to
18	the analogy that Taroon made, we know that
19	Medicare-insured adults over age 65 are at a
20	higher risk of readmission than other adults.
21	And so I think as you're pointing
22	out very well the population that you're

	Page 169
1	evaluating is a really important question.
2	And it's important to be careful to consider
3	who you compare.
4	But we don't feel that that limits
5	the ability to use a measure either in a
6	specific payer population or more broadly by
7	taking into account the fact that there are
8	differences among different payers.
9	CO-CHAIR KAPLAN: Okay, so I'm
10	hearing some confusion about that I think
11	probably although it does touch on feasibility
12	for certainly the data you've run, you've run
13	it. So you can get access to that data.
14	And I think the issue now is
15	shading over into usability and that's causing
16	a little bit of concern around here, although
17	the feasibility issue does anybody else
18	want to make a comment on feasibility? Are we
19	ready to vote? All right, let's vote.
20	MS. SHAHAB: Voting for
21	feasibility. One high, two moderate, three
22	low, four insufficient. Time begins now.

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	Page 170
1	We have all the votes for
2	feasibility. Three voted high, eighteen
3	moderate, one low and zero insufficient.
4	CO-CHAIR KAPLAN: Thank you. Now
5	we're into usability. And does anybody have
6	further comments on the potential usability of
7	this issue? Kathy, you want to go first?
8	DR. AUGER: I think it's just
9	worth noting the potential differences between
10	what is an unplanned readmission, an actual
11	preventable readmission.
12	I think there's a decent amount of
13	angst in the pediatric hospital medicine
14	community that there may not be a lot we can
15	do broadly to prevent readmission.
16	Although I would personally
17	comment it's still early in the game. It's
18	still a little hard to actually come down on
19	whether or not that's true or not.
20	There's certainly large groups
21	trying to reduce readmission rates nationally.
22	And frankly, the studies that they have done

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	Page 171
1	haven't yet been published so I think it's
2	still just it's just early to assess. But
3	I think it's a consideration is whether or not
4	how much this is actually preventable versus
5	planned. Unplanned, sorry.
6	CO-CHAIR KAPLAN: Paul?
7	DR. HEIDENREICH: Along similar
8	lines I do have significant concerns that
9	there's not a credible rationale for
10	improvement. I know looking through the
11	background that there have been almost no
12	studies of interventions but it doesn't seem
13	like there's either been or maybe you can
14	tell me if there are some studies that at
15	least correlating certain hospital practices
16	with better readmission rates even though
17	there wasn't an obvious randomization or
18	experimental design.
19	And without that it seems and
20	given that pediatrics I think is significantly
21	different from adult with multiple chronic
22	diseases, a large number of medications, it

	Page 172
1	would seem that your interventions you would
2	do would be different.
3	And so I'm very concerned that
4	it's not clear how hospitals will be able to
5	use the information once it's released.
6	CO-CHAIR KAPLAN: Developers want
7	to comment?
8	DR. NAKAMURA: Yes, Your Honor.
9	This is Mari Nakamura.
10	To start with Kathy's point about
11	preventability versus unplanned readmissions.
12	We agree that ideally it would be desirable to
13	be able to assess preventable readmissions
14	rather than unplanned readmissions.
15	We just don't think that that's
16	possible using claims data, using the data
17	that we currently have widely available to us.
18	And so we took the approach in harmony with
19	what has been done in adult measures to
20	instead focus on unplanned readmissions.
21	That said, research that we're
22	currently working on as well as conducted at

	Page 173
1	other institutions has indicated that a
2	sizeable proportion of unplanned readmissions
3	actually are potentially preventable.
4	Of course, how preventability is
5	determined is highly controversial. It's
6	something that I think will probably always be
7	a subject of debate because it's so
8	subjective.
9	We have a study that's still
10	currently underway in which we've talked with
11	patients, families, nurses, outpatient
12	doctors, inpatient doctors to try to get a
13	sense from all of them of how preventable they
14	thought readmissions were.
15	What we found is the more
16	information you get the better sense you get
17	that it's a very complicated question. And so
18	knowing that we are using claims data for this
19	measure we're choosing to focus on unplanned
20	readmissions.
21	Regarding the rationale for
22	whether readmission rates can be improved in

	Page 174
1	pediatrics it is indeed true that there are
2	far fewer studies in pediatrics than adults
3	looking at interventions for readmission.
4	There are a handful that have
5	focused on pediatric patients or that have
6	included pediatric patients and evaluated them
7	as a subset that have found, for example,
8	better adherence to practice guidelines has
9	been associated with reduced readmissions or
10	improvements in criteria for discharge.
11	All that said we acknowledge that
12	in this space for pediatric measures that more
13	likely what we're dealing with is a rationale
14	that makes sense rather than a lot of evidence
15	already in the literature.
16	And we believe that given that
17	processes such as discharge preparation and
18	education, making sure that there are good
19	transitions to the community are felt to be
20	equally important in pediatrics that
21	improvements in those processes could also
22	lead to improvements in readmission rates.

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	Page 175
1	The point that there are not as
2	many very chronically ill complex patients
3	among pediatric patients as adults is true.
4	That said, one of my colleagues at
5	our center conducted a study a couple of years
6	ago examining specifically the population of
7	patients with chronic complex conditions who
8	are frequently readmitted and found that while
9	they're a small percentage of all patients
10	they account for a high percentage of
11	readmissions in cost.
12	And so in terms of a population
13	that could be a focus for readmission
14	improvement that seems like a natural one.
15	Thank you.
16	CO-CHAIR KAPLAN: Thank you.
17	Taroon, do you have something specific to
18	this? Then let's go Leslie, Karen and Taroon.
19	MS. HALL: So, this is an
20	interesting area in usability and maybe should
21	be questioned or comments aimed towards
22	implementation.

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	Page 176
1	But communication and education
2	around this measure could not be stressed
3	more. Because readmission is something that
4	is known in the public I guess consciousness.
5	And now we apply the highly emotional area of
6	pediatrics. And without good explanation and
7	the fact that we have data sources coming from
8	Medicaid patients that were already felt that
9	are disenfranchised by other payer groups we
10	have the potential to have a good deal of
11	angst associated with this measure release.
12	And so I would just caution and
13	encourage that along with the implementation
14	guide we include communication plans and
15	education to the public.
16	CO-CHAIR KAPLAN: Certainly this
17	is one of those ones that if approved that MAP
18	is going to have something to say a lot about.
19	So, Karen?
20	DR. JOYNT: I just have two
21	additional thoughts about the usability. Both
22	again perhaps not things that would keep the

Page 177

1 measure from going forward but things that I think would be particularly essential in this 2 3 population to consider. And one is socioeconomic status. 4 I don't think you can get away from that in 5 6 this population. And especially because you've shown data that shows us that there's 7 a big difference there. Not necessary to 8 9 adjust for it but it's got to be part of the 10 discussion about how we think about this and 11 how it sort of rolls out. Because if we're just identifying hospitals that differ by 12 13 socioeconomic status I don't think we're doing 14 anyone a favor. And the second is administration 15 We've talked about this with the adult 16 rates. 17 measures as well. If a community works to reduce their admission rates for asthma they 18 could potentially increase their readmission 19 20 rates. And I think that would be equally true 21 for the complex chronic condition patients.

If you put a good intervention in place to 22

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	Page 178
1	improve their care as outpatients they might
2	not go into the hospital as much.
3	And I would hate to see hospitals
4	look worse on readmissions because they have
5	done such a good job of providing good
6	outpatient care for their patients.
7	I just don't think the long-term
8	quality here is just about readmission. It's
9	also got to be about admission. That's where
10	this stuff is moving. And I would really
11	encourage as we think about usability think
12	about some sort of companion way to examine
13	that piece.
14	CO-CHAIR KAPLAN: Thank you.
15	Taroon?
16	MR. AMIN: Just to follow up on
17	the question of the all-payer claims and the
18	MAX data.
19	I just wanted to clarify from the
20	developer is the information that's provided
21	in the testing form, that's a testing from the
22	MAX data set, right?

	Page 179
1	DR. NAKAMURA: This is Mari
2	Nakamura. Yes, that's right.
3	MR. AMIN: Is there information
4	that you can share with the committee around
5	the testing from the New York State all-payer
6	claims database that you've done?
7	DR. NAKAMURA: What we would be
8	able to provide is a sense of what the
9	readmission rates look like, how they compared
10	to the Medicaid readmission rates.
11	We also specifically used our all-
12	payer data sets to be able to evaluate some
13	questions that we couldn't in MAX. So,
14	specifically insurance status. And also the
15	relationship between insurance status and
16	race/ethnicity and risk factors for
17	readmission.
18	That said, we did not do all of
19	the testing in the all-payer data set that we
20	did with the MAX data set in part because we
21	were limited in the all-payer data set to not
22	as many states and we felt that the MAX data

	Page 180
1	set was actually far larger. And although it
2	was limited to Medicaid patients based on the
3	power that we had the more desirable data set
4	to use as our primary development data set.
5	CO-CHAIR KAPLAN: Thank you.
6	We're over time for this measure by a
7	considerable amount.
8	On the other hand, since this is
9	the first of its kind I feel like this was a
10	very good discussion of the kinds of issues
11	that are problematic.
12	With respect to Medicaid status
13	and Karen's issue about socioeconomic status,
14	if you use it as an adjuster you're adjusting
15	away the thing you're trying to explain.
16	So it's one of these problematic
17	areas that once again we are faced with here's
18	new ground and we're faced with the measure we
19	have in front of us, not the one we would like
20	to develop 5-10 years hence. So, staring at
21	the thing you're staring at, this issue we're
22	now preparing to vote for. Any other
	Page 181
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1	comments?
2	CO-CHAIR HALL: I do. I'm not an
3	expert in gates data and I'm still not sure
4	what I've heard. So a couple of the experts
5	maybe could comment.
6	The extrapolatability from
7	Medicaid to the general population sounded
8	like there were concerns about the resolved
9	in my head.
10	Those still exist. Anybody want
11	to?
12	DR. FIELDS: Yes, I think the
13	tough thing about this, I'm just trying to
14	summarize what I've heard and what I know is
15	that ideally the first measure for pediatric
16	care would probably be more related to access
17	to care and especially primary care and
18	surveillance. It wouldn't be about hospital
19	services. Children have the lowest overall
20	admission rate of any cohort seen in the
21	emergency department, for example, in terms of
22	unscheduled admissions or readmissions.

	Page 182
1	So, in a perfect policy
2	development world you wouldn't start with the
3	hospital piece. You'd probably end with it.
4	The other fundamental problem
5	which we touched on yesterday is that the
6	place you need to move the rock with the most
7	urgency are the very states where you're going
8	to have the poorest Medicaid data and the
9	smallest number of children who should be
10	getting better primary care eligible for
11	Medicaid services.
12	So it's a fundamental conundrum.
13	Because as with other applications of
14	administrative data this is an easy place to
15	start. It may be the only place to start, but
16	it's not the best place to start.
17	CO-CHAIR KAPLAN: Thanks. Kathy?
18	DR. AUGER: I would comment that I
19	agree with you that certainly some aspects of
20	pediatric readmission are hinge upon things
21	like access to care although I don't think
22	that that's all of it.

	Page 183
1	And certainly I think in terms of
2	I think it's a real question of whether or
3	not hospitals can truly move this metric
4	although it seems like with the new metric the
5	standard is to reassess in three years and see
6	whether or not if there was any change in
7	time.
8	But then I would just also I
9	think to me the bigger question is what Bruce
10	raised. It's primarily based on Medicaid data
11	and how it extracts beyond Medicaid is still
12	a little bit of a question mark in my mind.
13	CO-CHAIR KAPLAN: I would just add
14	from a non-physician, non-provider standpoint
15	it isn't a good idea if children are
16	readmitted to the hospital. So the idea that
17	readmission to the hospital is a quality
18	measure is not a stretch for me.
19	I mean, the question is rather as
20	Bruce alluded to who are we extrapolating to
21	from who are we actually able to measure this
22	on. And that remains a difficulty.

Page 184 1 On the other hand we are again, right Taroon? Stuck with -- we're not stuck 2 with --3 (Laughter) 4 CO-CHAIR KAPLAN: -- we have what 5 exactly we have in front of us and that's what 6 we're considering. 7 8 CO-CHAIR HALL: We're looking where the light is. Leslie. 9 10 MS. HALL: I guess my concern 11 about the Medicaid data is somewhat helped by the thought that if we help the poor we help 12 13 everyone. CO-CHAIR KAPLAN: Okay. Go ahead. 14 MS. TRAVIS: I guess -- this is 15 just a clarification because I might have 16 17 gotten confused along the way. If this measure is fully tested in 18 the Medicaid population with the validity 19 20 testing and not fully tested in the all-payer 21 or in other settings, when we approve it it will say it was tested in the Medicaid 22

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	Page 185
1	population how does that affect how it is used
2	out of the portfolio?
3	Because usually if I remember
4	correctly we've always said that it's endorsed
5	based on how it's specified and how it's
6	tested I think was the second part of that.
7	So I'm just trying to ask for clarification on
8	that.
9	CO-CHAIR HALL: It's not specified
10	to be exclusively used in Medicaid. It is
11	the details reveal that it was tested on
12	Medicaid so my interpretation is that it would
13	not be limited to application in Medicaid.
14	But then in follow-up somebody would see how
15	it performs. Taroon, do you want to add to
16	that?
17	MR. AMIN: Yes, and I think that
18	seems to be the conundrum that you've raised
19	which started this conversation which is that
20	it's not specified the testing doesn't
21	match completely the specifications.
22	So if it were to be used outside

	Page 186
1	of Medicaid, you know, it's specified to be
2	able to do that. Although the testing doesn't
3	demonstrate how it would perform outside of
4	Medicaid.
5	CO-CHAIR KAPLAN: So, are we close
6	to ready? It's my understanding that if you
7	did nothing else you would overestimate the
8	amount of the readmission rates by
9	hospital. And then the question is whether or
10	not that overestimation would compromise its
11	usability for when measures got less and less
12	stable because the numbers got smaller and
13	smaller for readmission rates. So, that could
14	be addressed in a use if approved.
15	DR. NAKAMURA: Thank you. This is
16	Mari Nakamura. You're correct that
17	readmission rates will tend to be higher in a
18	Medicaid-only population we found than in an
19	all-payer population.
20	There's some different sort of
21	considerations that work in going from a
22	Medicaid-only to an all-payer data set.

	Page 187
1	So, for example, we would
2	anticipate that for many hospitals the sample
3	sizes used would actually get better because
4	while Medicaid is a sizable portion of all
5	hospitalizations it's only about one-third.
6	And of course that will differ depending on
7	the proportion of Medicaid at a given
8	hospital.
9	But if anything we would expect
10	that reliability would improve and that the
11	position of rates would actually get better.
12	CO-CHAIR KAPLAN: Gotcha. The
13	numerator would okay.
14	CO-CHAIR HALL: I think the
15	numerator would get better but fundamentally
16	you are you're drawing coefficients out of
17	a higher-risk population which means you're
18	less likely to hold a hospital accountable
19	that has a lower risk population.
20	DR. NAKAMURA: This is Mari
21	Nakamura. Thank you for that question because
22	I hadn't realized I apologize part of

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Page 188 1 the concern here. So the way that our measure is run 2 is that actually we don't provide the beta 3 coefficients as set numbers to be used in the 4 Instead the model is actually run on model. 5 the data set to which it's being applied and 6 new coefficients for that particular 7 8 population are generated. 9 So I agree that the choice of 10 covariates we made certainly was based 11 primarily on the Medicaid data set. We did test the very same model on our all-payer data 12 13 set and we believe that in terms of the relationships between things like age and 14 gender and chronic conditions that one can 15 16 generalize from the Medicaid population to an 17 all-payer data set in terms of those fixed effects relationships while recognizing that 18 at the same time having Medicaid insurance as 19 20 a child is, it appears, an additional risk factor on top of those patient 21 characteristics. 22

Page 189 1 CO-CHAIR HALL: Thank you. CO-CHAIR KAPLAN: Fair enough. 2 3 Kathy. DR. AUGER: Do you have a sense of 4 what the C statistic was in the all-payer 5 model? Is it the same? 6 DR. NAKAMURA: This is Mari 7 Nakamura. No, I'm sorry, I don't have that 8 9 although we would be able to easily determine 10 that. 11 CO-CHAIR KAPLAN: Okay. I think we are ready to vote on the issue of, where 12 13 are we, usability? Usability. MS. SHAHAB: Voting for usability 14 and use. One is high, two moderate, three 15 16 low, four insufficient information and time 17 starts now. We have all the votes for 18 usability and use. Zero high, fourteen 19 20 moderate, eight low, zero insufficient 21 information. 22 CO-CHAIR KAPLAN: Thank you. And

	Page 190
1	drum roll, drum roll, we are now to
2	endorsement. Any more further thoughts?
3	Okay.
4	MS. SHAHAB: Voting for overall
5	suitability for endorsement. One is yes, two
6	no. Time begins now.
7	We have all the votes for overall
8	suitability for endorsement. Measure 2393
9	Pediatric All-Condition Readmission Measure,
10	the votes are 17 yes, 5 no.
11	CO-CHAIR KAPLAN: Thank you very
12	much and thank you to the developers. Now we
13	are 20 minutes behind. And so we are going to
14	have to sort of make tracks.
15	I assume much of the discussion
16	will now sharpen up and condense itself around
17	many of either not similar issues but possibly
18	easier to deal with issues.
19	So, would you reintroduce
20	yourselves for the record and then describe
21	the measure.
22	DR. NAKAMURA: Thank you. This is

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	Page 191
1	Mari Nakamura from Boston Children's Hospital.
2	DR. ZASLAVSKY: Alan Zaslavsky
3	from Harvard Medical School.
4	DR. NAKAMURA: And we have Mark
5	Schuster on the phone. I will keep my
6	introduction brief because by design these two
7	measures are very similar and so for the most
8	part the same considerations apply.
9	Our pediatric lower respiratory
10	infection measure evaluates readmissions
11	following an index hospitalization for
12	bronchiolitis, influenza, or community-
13	acquired pneumonia.
14	We decided on a measure focusing
15	on lower respiratory infections because
16	they're among the most common reasons for
17	hospitalization in children.
18	In addition, they're among the
19	diagnoses with the most prevalent
20	readmissions. We found an overall 30-day
21	readmission rate of 5.6 percent which
22	corresponds with the large absolute number of

	Page 192
1	readmissions given the high number of initial
2	hospitalizations for lower respiratory
3	infection.
4	We prioritize harmonizing this
5	measure with NQF-endorsed adult readmission
6	measures as well as with our all-condition
7	measure. And as a result the approaches we
8	used in developing the measure are very
9	similar.
10	We used the same case mix
11	adjustment model because we found that it
12	performs very well for LRI readmissions with
13	regard to discrimination and calibration.
14	The issue of limited reliability
15	due to small sample sizes is even more of an
16	issue for any condition-specific rate and
17	this is true for LRI than for all condition
18	rates. But again we found that a majority of
19	children were cared for at higher-volume
20	hospitals than as a result were at hospitals
21	with good reliability.
22	The New York Office of Safety and

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	Page 193
1	Quality also tested this measure on its
2	Medicaid and all-payer databases and noted
3	that having implemented one measure it
4	required minimal effort to be able to
5	implement the other.
6	And so we do think that this
7	measure could be a useful tool to evaluate
8	quality and encourage improvements in care for
9	an important pediatric condition. Thank you.
10	CO-CHAIR KAPLAN: Thank you. I
11	think Mari is the only person in the room that
12	talks faster than I do so thank you very much
13	for that quick summary.
14	Jo Ann, do you want to talk to us
15	about evidence?
16	DR. BROOKS: I'll go ahead and get
17	started with this.
18	As was said this is a companion
19	measure to the one we just discussed. And
20	looking at lower respiratory infections,
21	accounting for a large number of the
22	readmissions we see it is an important area.

	Page 194
1	And there's good support here on
2	how these readmissions, we may be able to look
3	at things to improve the readmissions, looking
4	at our key processes, our discharge planning,
5	care transitions, appropriate follow-up, et
6	cetera.
7	And disparities exist for many of
8	these differences we see in patients being
9	readmitted for lower respiratory infection.
10	CO-CHAIR KAPLAN: Thank you.
11	Kathy, nothing to add? Others? Vote
12	evidence.
13	MS. SHAHAB: Voting for 1(a)
14	evidence, one yes, two no. Time begins now.
15	We have all the votes for 1(a)
16	evidence. Nineteen voted yes, two voted no.
17	CO-CHAIR KAPLAN: Performance gap.
18	Jo Ann?
19	DR. BROOKS: There exists a
20	performance gap for this. This is also a
21	newly commissioned measure by CMS and AHRQ.
22	There's disparities in care across populations

	Page 195
1	in many different ways and there are strong
2	data to support that there is a quality gap
3	and a need for this measure.
4	There is support in the
5	application talking about some of the
6	pediatric data that's out there although it is
7	not as rich as the adult data. But the
8	appropriate rationale is there.
9	CO-CHAIR KAPLAN: Kathy?
10	DR. AUGER: I'd just comment
11	though that the readmission rate for the lower
12	respiratory tract infections is actually lower
13	than the all-cause.
14	And so it's hard to know what the
15	like how much of a range we're actually
16	dealing with at the different hospitals, like
17	what the interquartile range would be for the
18	risk-standardized rate and whether or not
19	that's significant.
20	CO-CHAIR KAPLAN: Do the
21	developers have any information for us?
22	DR. NAKAMURA: This is Mari

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	Page 196
1	Nakamura. I guess thank you.
2	So, for our lower respiratory
3	infection measure to give you analogous
4	numbers to what I provided for all-condition,
5	a hospital that's two standard deviations
6	below the mean would have a readmission rate
7	of 1.7 percent.
8	A hospital two standard deviations
9	above would have a readmission rate of 12.6
10	percent. So, there's actually a wider range
11	for lower respiratory infection.
12	CO-CHAIR KAPLAN: Thank you for
13	that. Others? Okay, ready to vote?
14	Performance gap.
15	MS. SHAHAB: Voting for 1(b)
16	performance gap. One high, two moderate,
17	three low, four insufficient. And the time
18	begins now.
19	We have all the votes for
20	performance gap. Three voted high, eighteen
21	moderate, zero low and zero insufficient.
22	CO-CHAIR KAPLAN: Thank you.

Page 197 1 Priority. Jo Ann? DR. BROOKS: And this continues to 2 3 be a high-priority measure as it was commissioned by AHRQ and CMS. Also because it 4 relates and impacts a large number of 5 pediatric patients and accounts for a large 6 number of the readmissions in hospitals. 7 8 CO-CHAIR KAPLAN: Thank you. 9 Kathy? 10 DR. AUGER: Yes, just lower 11 respiratory tract infections are one of the most common indications for hospitalization in 12 13 pediatrics so to me it's high priority. CO-CHAIR KAPLAN: Others? Ready 14 to vote? 15 MS. SHAHAB: Voting for 1(c) high 16 17 priority. One high, two moderate, three low, four insufficient. Time begins now. Just one 18 19 more vote. We have all the votes for 1(c) 20 21 high priority. Twelve voted high, eight moderate, one low and zero insufficient. 22

	Page 198
1	CO-CHAIR KAPLAN: Thank you very
2	much. Scientific acceptability. Start with
3	reliability. Jo Ann?
4	DR. BROOKS: Yes, this measure was
5	tested the exact same way as the previous
6	measure using the MAX data. And when we
7	looked at reliability it ranged between 0.5 to
8	0.77.
9	The measure was also found to be
10	highly reliable at hospitals with an adequate
11	sample size, but obviously it did not perform
12	as well in those with lower sample size.
13	And one of the questions was will
14	exclusions make the reliability across time
15	and place an issue for this measure. Because
16	there's a large number of exclusions in this
17	measure.
18	DR. NAKAMURA: This is Mari
19	Nakamura. Regarding exclusions the one
20	difference with our all-condition measure is
21	the case definition requirement for the index
22	hospitalizations.

	Page 199
1	So, speaking more generally about
2	both measures I think that we've created an
3	appearance of maybe more exclusions than
4	typical because we listed everything in terms
5	of being exclusion rather than some as
6	inclusions.
7	So, some of these are, for
8	example, based on limiting to the pediatric
9	age range. Some other common exclusions that
10	we saw have been used quite uniformly in adult
11	measures such as excluding patients who leave
12	AMA or certainly patients who die in the
13	hospital.
14	Our other set for the exclusions
15	that are quite similar to the adult
16	readmission measures are certain data quality
17	exclusions for key variables with missing or
18	what looked like poor quality data. For
19	example, a discharge date that occurs before
20	a date of birth.
21	The main clinical exclusions that
22	we apply for both measures are, first of all,

Page 200
patients who are receiving obstetric care with
the rationale that these patients are
typically not under the purview of pediatrics
even if they fall in the pediatric age range.
And so we felt would most likely be better
included in obstetric measures rather than a
pediatric measure.
We also exclude patients with a
primary mental health diagnosis. That is
consistent with other measures but in our own
testing as well we felt it was justified
because we found that readmission people for
a given hospital does not track for mental
health conditions as compared with other
conditions.
And then finally we exclude
newborns who are in the hospital for their
birth admission. The clinical rationale for
that is that they are among all patients in
the hospital not actually there because
they're ill, but rather for another life
event. And so we felt that it made sense not

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	Page 201
1	to include them in a readmission rate.
2	CO-CHAIR KAPLAN: Thank you. I'm
3	going to ask the committee to hold that
4	consideration for the validity discussion as
5	opposed to unless those exclusions change
6	the reliability we need to keep focused on
7	reliability. Kathy?
8	DR. AUGER: I would just comment
9	as Mari already acknowledged the reliability
10	in this metric is not quite as good as the
11	previous measure in that only 229 of the 1,743
12	hospitals actually had a readmission rate
13	reliability of greater than 0.5.
14	But that again these hospitals
15	accounted for 62 percent of the lower
16	respiratory tract infection hospitalizations.
17	So it's still even though it's a smaller
18	number of hospitals it's still a majority of
19	the hospitalizations.
20	CO-CHAIR KAPLAN: The precision of
21	those estimates once again sort of tracks with
22	everything else we're seeing. In fact, it

	Page 202
1	lands a little bit on the higher side compared
2	to other reliability estimates we've been
3	seeing in some of these other measures.
4	Anybody else? Vote reliability.
5	MS. SHAHAB: Voting for 2(a)
6	reliability. One is high, two moderate, three
7	low, four insufficient. Time begins now.
8	We're still waiting on two more votes. One
9	more.
10	All votes are in for 2(a)
11	reliability. The results are 1 high, 18
12	moderate, 2 low and zero insufficient.
13	CO-CHAIR KAPLAN: Thank you.
14	Validity. And I'll ask you to keep in mind
15	the exclusion discussion we just had.
16	DR. BROOKS: On the validity, the
17	construct validity was demonstrated as
18	associated with the literature and the quality
19	of care. The processes related to reductions
20	and readmissions that we're all aware of.
21	Criterion validity was shown using
22	a data set from Boston Children's Hospital

	Page 203
1	over a 1-year period. And the sensitivity and
2	the specificity were 87.0 and 99.7 percent
3	respectively.
4	They also did face validity on the
5	planned procedure algorithm that was completed
6	and also received public comments from the
7	Federal Register. And they then took those
8	comments and put those into the methodology to
9	improve it.
10	And those would be my comments.
11	DR. AUGER: So, very similar
12	issues with validity in terms of identifying
13	unplanned readmissions as appropriate.
14	The couple of questions well,
15	so I'd say that the model calibration is good
16	and the C statistic for predictive ability was
17	0.71 so that's in a reasonable range.
18	The one question that we were just
19	talking about is how CF exacerbations come
20	into play here, whether or not the model would
21	adequately account for CF as well.
22	DR. NAKAMURA: This is Mari

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Page 204 1 Nakamura. So, cystic fibrosis in particular 2 would end up being included in the model as 3 one of the chronic condition indicator 4 variables for chronic respiratory infections. 5 6 But this is also an opportunity --I'll keep it very brief -- to mention that we 7 thought about whether we needed to have more 8 symptom variables for chronic conditions for 9 10 the LRI measure versus the all-condition. 11 And in reflecting decided not to because it's actually a wide range of 12 13 conditions that place patients at higher risk of severe lower respiratory infections. 14 They're not just respiratory, but for example, 15 cardiac, neurological. And so we felt that it 16 17 made sense to keep all of the chronic condition indicators and found indeed that the 18 model actually performs very well. 19 20 CO-CHAIR KAPLAN: Thank you. Ι want to ask for a point of clarification about 21 criterion validity because that implies a gold 22

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	Page 205
1	standard. You weren't using Boston Children's
2	as the criterion. Did I misunderstand that?
3	How is what you did criterion validity?
4	DR. NAKAMURA: This is Mari
5	Nakamura.
6	So we were using the electronic
7	health record data as our gold standard based
8	on having performed detailed chart reviews of
9	the cases that we evaluated with the idea that
10	such a chart review is at least a better
11	standard than what can be found from claims
12	data in terms of being able to assess both
13	whether readmission occurred as well as
14	whether it met our definition for an eligible
15	readmission meaning that it wasn't for a
16	planned procedure or chemotherapy.
17	CO-CHAIR KAPLAN: Thank you. Some
18	of us would call that convergent validity
19	because it converges with a different data
20	source as opposed to criterion validity. But
21	that's okay.
22	DR. NAKAMURA: Thank you.

Page 206 1 CO-CHAIR KAPLAN: Other comments on validity? Let's vote. 2 MS. SHAHAB: 2(b) validity. 3 One 4 is high, two moderate, three low, four 5 insufficient. Time begins now. One more 6 vote. All votes are in for 2(b) 7 8 validity. Zero voted high, twenty moderate, 9 one low and zero insufficient. 10 CO-CHAIR KAPLAN: Thank you. 11 Feasibility? DR. BROOKS: And on feasibility I 12 would say that since it's claims data it's 13 14 easily feasible for us to get these data. The one question -- some of the 15 16 concerns we've already had is it's based on 17 Medicaid data. But is it feasible as claims data. 18 CO-CHAIR KAPLAN: Others? 19 So 20 we're burned out on feasibility. 21 (Laughter) CO-CHAIR KAPLAN: Ready to vote. 22

	Page 207
1	MS. SHAHAB: Voting on
2	feasibility. One high, two moderate, three
3	low, four insufficient. Time begins now.
4	Still waiting on two more votes.
5	All votes are in for feasibility.
6	The results are 3 high, 17 moderate, 1 low and
7	zero insufficient.
8	CO-CHAIR KAPLAN: Thank you. Onto
9	usability.
10	DR. BROOKS: Usability, since this
11	measure is really a subset of the previous
12	measure all the issues we've discussed with
13	usability in the previous discussion are the
14	same here.
15	CO-CHAIR KAPLAN: Kathy?
16	DR. AUGER: So yes, again, just
17	the whole issue of preventability versus
18	unplanned is a consideration. But that's the
19	same as the other metric.
20	The one thing that I would comment
21	might be different for this metric compared to
22	the other metric is this is lower

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	Page 208
1	respiratory infection is a very seasonal
2	illness. And so therefore as it's written I
3	think it's not an issue because it's an annual
4	evaluation.
5	But it's something that it would
6	not be appropriate for it to be used as a
7	quarterly evaluation because of the seasonal
8	variability in this. But I see it as written
9	as fine.
10	CO-CHAIR KAPLAN: Want to respond
11	to that?
12	DR. NAKAMURA: Thank you. This is
13	Mari Nakamura. Kathy is correct that if we
14	were to try to report these results as
15	quarterly that season would be a really
16	important consideration.
17	We chose to make it annual, taking
18	care of that fact but also recognizing that
19	quarterly rates would be an even greater
20	challenge with regard to sample size.
21	CO-CHAIR KAPLAN: Thank you.
22	Other comments? Ready to vote.

	Page 209
1	MS. SHAHAB: Voting on usability
2	and use. One high, two moderate, three low,
3	four insufficient in. The time begins now.
4	All votes are in for usability and
5	use. The results are zero high, 17 moderate,
6	4 low and zero insufficient information.
7	CO-CHAIR KAPLAN: Are we ready to
8	vote on endorsement? Any discussion?
9	DR. BROOKS: My only comment is
10	that I think this one is very specific to
11	sociodemographic data as well and we need to
12	consider that as we move forward with this
13	down the road.
14	CO-CHAIR KAPLAN: Yes, the issue
15	of sociodemographic data as we've discussed
16	yesterday and today is going to be, you know,
17	one of these things that is going to be guided
18	by a committee that hasn't yet kind of given
19	its guidance to us. So we are going to either
20	endorse or not endorse the right.
21	CO-CHAIR HALL: Understanding that
22	we can all as a group express that we want our

	Page 210
1	NQF colleagues to capture that for these two
2	measures the group in particular felt very
3	strongly that this question needs to be
4	addressed in the future.
5	Do people feel that's a relative
6	consensus? Anyone would object to attaching
7	that comment?
8	CO-CHAIR KAPLAN: No, but and the
9	issue of Medicaid and SES adjustment is one of
10	these things that's also going to present a
11	rather dodgy problem. Because socioeconomic
12	status and some of the adult measures actually
13	use Medicaid status as a proxy for
14	socioeconomic status. So it's going to be a
15	more complex issue with some of these
16	measures.
17	CO-CHAIR HALL: All I'm suggesting
18	is we create a bit of a flag for our NQF
19	colleagues that if someday comes where the
20	white paper recommendations have changed on
21	this that this is an easy flag to spot.
22	CO-CHAIR KAPLAN: So flaggage

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	Page 211
1	approved and associate recommendations.
2	(Laughter)
3	CO-CHAIR KAPLAN: Are we ready?
4	Any other comments before we vote for
5	endorsement or non-endorsement?
6	MS. SHAHAB: Voting for overall
7	suitability for endorsement. One yes, two no.
8	Time starts now.
9	All votes are in for overall
10	suitability for endorsement for measure 2414
11	Pediatric Lower Respiratory Infection
12	Readmission Measure. The results are 18 yes
13	and 3 no.
14	CO-CHAIR KAPLAN: Thank you very
15	much. And now we are ready for NQF member and
16	public comment. And we will invite public
17	comment. Thank you to the developers for a
18	nice summary and discussion.
19	We're onto the NQF member and
20	public comment. I'm going to turn over to
21	Adeela.
22	MS. KHAN: Kathy, can we compile

Page 212 1 the list, please? **OPERATOR:** Yes, ma'am. If you 2 would like to make a public comment please 3 press \* then the number 1. No, no public 4 comments at this time. 5 6 MS. KHAN: Do we have any public comment in the room? 7 8 DR. SCHWALENSTOCKER: Good 9 afternoon -- or I guess it's good morning. 10 And I realize I'm standing between you and 11 lunch. My name is Ellen Schwalenstocker. 12 13 I'm with the Children's Hospital Association. 14 And I just wanted to highlight 15 some of the discussion among the committee 16 that I just think is really important some of 17 which is outside your purview. But we've been saying for a long 18 time that it's really important to develop 19 20 good pediatric measures. And I want to thank the Center of Excellence at Boston Children's 21 Hospital for their work in developing these 22

Page 213 1 measures. I'm not sure, and this may be a 2 3 question for you, what the 3-year experience sort of guidelines are. I do strongly agree 4 with the recommendation on should the NOF 5 6 change its policy on adjustment for SES and other factors, that it would be really 7 important to look at these measures as well as 8 9 adult readmissions measures in that light. 10 I also think because of the lack of a large set of good pediatric measures it 11 will really be important to have other 12 13 measures available before these kinds of measures are used for, say, accountability 14 purposes or pay-for-performance. 15 And so I'm wondering if that's 16 17 sort of part of the 3-year process to give us more chance for validation of measures against 18 other measures? I guess that's convergent 19 20 validity. And I know there are some other 21 measures that will be coming forward from the 22

	Page 214
1	CHIPRA Centers of Excellence which we just
2	think is a really important program and are
3	glad to see the measures coming to NQF.
4	MR. AMIN: So, I can address that,
5	Ellen, right now.
6	So, the NQF reevaluates all of its
7	measures that are recommended for endorsement
8	in a 3-year maintenance cycle.
9	In that 3-year maintenance cycle
10	we look for experience on the measure. In
11	particular in the importance to measure
12	criteria we're looking for some actual, at the
13	measure performance gap information we're not
14	just looking conceptually whether there's a
15	measure performance gap, we want to see the
16	performance gap in the measure itself and to
17	see the distribution and if there's any
18	overall gap in performance or overall less
19	than optimal performance.
20	And then also in the use and
21	usability criteria we want to see if the
22	measure has been implemented for use in

	Page 215
1	quality improvement applications and then a
2	further time line is ready for accountability
3	applications which we include public reporting
4	in that domain.
5	So the purpose there is that we
6	have measures that are actually picked up by
7	the field and we can demonstrate that they're
8	actually being used for the purposes of
9	quality improvement and reporting the
10	information to the public.
11	DR. BURSTIN: I'm just wanting to
12	add again if this measure is endorsed and goes
13	through the entire process, again, the SES
14	issue I think outstanding. We don't know how
15	that's all going to play out.
16	But I think certainly we'll have
17	to make a decision with our CSAC and our board
18	depending on how that lands in terms of
19	whether we'll bring measures back sooner than
20	the 3-year time frame for maintenance if in
21	fact there are additional issues to address.
22	But you know, a very important

Page 216
piece of this. As these measures are going
out there NQF endorsement does imply they are
appropriate for any accountability application
and could get picked up for any of those
certainly by CMS or others.
We would really encourage the
field to be really vigilant about keeping
track of measures that are really helping to
drive improvement, measures for which there
may be unintended consequences and bring that
forward, that information, in realtime rather
than just waiting for the 3-year limit.
MR. AMIN: Are there any other
public comments?
OPERATOR: Once again to make a
public comment please press *1. You have a
comment from John Muldoon with 3M Health
Information Systems.
MR. MULDOON: Are you able to hear
me?
MS. KHAN: Yes, we can hear you.
MR. MULDOON: Several comments.
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	Page 218
1	Also, when you go out to 30 days
2	and the number of unrelated and low
3	preventability readmissions increases. So
4	that compounds the concern and the approach to
5	readmissions.
6	On the risk adjustment concern
7	that often the principal diagnosis and the
8	acuity of the admission and really complex
9	prior conditions that have a very big impact
10	on admission rates and readmission rates.
11	And that's not very specifically
12	addressed. It's more of a generic approach to
13	any chronic condition from a list of about
14	4,500, mild, moderate and severe, many of
15	which have very little influence over
16	admissions and readmissions.
17	So, we're concerned that that's
18	not picking up the readmission factors very
19	well.
20	And in terms of testing and
21	evaluation I don't think we really saw it
22	tested across different subgroups of pediatric

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	Page 219
1	patients and across different hospitals who
2	serve different populations.
3	So I think there's a lot that we
4	don't know and need to be concerned about as
5	it gets rolled out if it's endorsed.
6	And just to illustrate, for lower
7	respiratory infections for the pediatric
8	population those with major chronic
9	conditions.
10	We've done a lot of research and
11	analysis on this such as cystic fibrosis as
12	the committee discussed, bronchopulmonary
13	dysplasia, ventilator-dependent patients,
14	chronic respiratory failure. They tend to
15	have readmission rates in the 10 to 15 percent
16	range compared to otherwise healthy children
17	in the 3 to 5 percent range.
18	And we just don't see how that can
19	be teased out with the more generic risk
20	adjustment methods. So those are the thoughts
21	that we'd like to share.
22	And if the committee does endorse

	Page 220
1	I think there's a lot of cautions to go
2	forward. Thank you.
3	MR. AMIN: Thank you. Is there
4	any other comments on the phone?
5	OPERATOR: At this time there are
6	no public comments on the phone line.
7	MR. AMIN: There are no other
8	public comments in the room.
9	CO-CHAIR HALL: I think we can
10	break for lunch then. Just a quick note for
11	those who weren't here yesterday. We would
12	ask the audience members to wait for the
13	committee members to grab their lunch first
14	before helping yourself.
15	CO-CHAIR KAPLAN: So we'll
16	reconvene at 12:15.
17	(Whereupon, the foregoing matter
18	went off the record at 11:41 a.m. and went
19	back on the record at 12:13 p.m.)
20	CO-CHAIR HALL: We have three
21	measures left this afternoon, one from the
22	American College of Cardiology and two from

	Page 221
1	Yale CMS.
2	And then we have a very important
3	brief update on measure 1789. So we're hoping
4	we'll still have faces around the table when
5	we get to that point.
6	We'll ask the American College of
7	Cardiology representatives to introduce
8	themselves and briefly introduce their
9	measure, please.
10	DR. CURTIS: Hi, this is Jeptha
11	Curtis from Yale also representing the ACC
12	today.
13	CO-CHAIR HALL: Hang on, I'm not
14	sure we heard that.
15	DR. CURTIS: Sorry. Jeptha Curtis
16	from Yale representing the ACC today.
17	MS. SLATTERY: So Dr. Curtis will
18	be speaking primarily to the measure
19	methodology but I'm Lara Slattery, ACC senior
20	director for scientific reporting and I can
21	address any questions related to
22	implementation.

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	Page 222
1	DR. CURTIS: So, for endorsement
2	maintenance today under review is the Hospital
3	30-day Risk-standardized Readmission Rates
4	Following Percutaneous Coronary Intervention.
5	It is a measure that was endorsed
6	I believe 3 or 4 years ago. And identifies
7	unplanned readmission rates for hospitals that
8	perform percutaneous coronary interventions on
9	Medicare fee-for-service patients greater than
10	65 years old.
11	As I mentioned this is unplanned
12	readmissions and we updated the measure from
13	the initial endorsement to include a modified
14	version of the hospital-wide readmission
15	algorithm that identifies unplanned
16	readmissions. Basically including a larger
17	number or considering a larger number of
18	readmissions planned.
19	In addition, we've updated the
20	measure to include the use of direct as
21	opposed to indirect identifiers to match the
22	CathPCI registry data with CMS information

	Page 223
1	about readmission which is used to identify
2	the risk-standardized rates.
3	And has also in the interim
4	actually gone into implementation. And last
5	March the hospitals received their reports as
6	to their risk-standardized readmission rates.
7	And then there was a voluntary public
8	reporting of those rates on Hospital Compare
9	as well as the ACC's internal websites.
10	So that this information has or
11	this measure has progressed from development
12	to approval to implementation in a relatively
13	short time frame. And I think it's open for
14	questions.
15	CO-CHAIR HALL: Thank you very
16	much. We will start with the category of
17	evidence and our lead discussants are Mae and
18	Larry. So I invite them to kick it off.
19	DR. GLANCE: I'll go ahead and
20	start. In terms of evidence to support the
21	measure focus the measure developers present
22	evidence that there is a high level of

Page 224 1 variation across hospitals which to me represents strong evidence. 2 When you have variation in 3 outcomes that means that there's the potential 4 to improve your outcomes assuming that you 5 have properly adjusted for differences in case 6 mix. So I think the evidence is strong. 7 CO-CHAIR HALL: Mae said she 8 9 agreed but I didn't -- is your mike working? 10 MS. CENTENO: I agree. 11 (Laughter) CO-CHAIR HALL: Okay, great. Any 12 13 other comments on evidence? Okay, we'll move 14 to vote. MS. SHAHAB: Voting for 1(a) 15 16 evidence. One is yes, two is no and the time begins now. 17 18 Twenty yes and zero no. CO-CHAIR HALL: Moving into 19 20 performance gap, opportunity. 21 DR. GLANCE: In terms of the performance gap without risk adjustment the 22

Page 225 1 difference between the lowest decile and the highest decile was zero percent readmission 2 3 rates versus 28 percent readmission. After risk adjustment the 4 5 difference between the 10th percentile and 6 90th percentile was 13.5 versus 10.1 percent. So again, evidence of a significant 7 performance gap. 8 9 CO-CHAIR HALL: Additional 10 comments? Okay. 11 MS. SHAHAB: Voting for 1(b) performance gap. One high, two moderate, 12 13 three low, four insufficient and your time 14 starts now. 15 All votes are in for performance 16 Seventeen high, four moderate, zero low gap. 17 and zero insufficient. CO-CHAIR HALL: Priority. 18 19 DR. GLANCE: In terms of priority 20 this is one of the conditions that was 21 identified as a high-priority condition by MedPAC. 22

	Page 226
1	The overall incidence of
2	readmissions within 15 days was 10 percent
3	which represents roughly about 44,000
4	readmissions in 2005 at a cost of \$360 million
5	annually. So, I would suggest that this is a
6	high-priority condition.
7	CO-CHAIR HALL: Additional
8	comments? Seeing none.
9	MS. SHAHAB: Voting for 1(c) high
10	priority. One is high, two moderate, three
11	low, four insufficient and time begins now.
12	Three more votes.
13	All votes are in for 1(c) high
14	priority. Eighteen high, three moderate, zero
15	low, zero insufficient.
16	CO-CHAIR HALL: Scientific now.
17	Reliability and validity.
18	DR. GLANCE: In terms of
19	reliability data reliability. This is
20	based on clinical data which is audited using
21	annual onsite chart reviews and data
22	abstraction.

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	Page 227
1	In terms of model reliability the
2	intraclass correlation coefficient was 0.37
3	which is indicative of fair agreement and very
4	much in the zone of the other measures that we
5	have looked at over the past two days.
6	CO-CHAIR HALL: Additional
7	comments? Seeing none.
8	MS. SHAHAB: Voting for 2(a)
9	reliability. One high, two moderate, three
10	low, four insufficient and the time starts
11	now. One more vote.
12	All votes are in for 2(a)
13	reliability. Five high, sixteen moderate,
14	zero low and zero insufficient.
15	CO-CHAIR HALL: Validity.
16	DR. GLANCE: In terms of validity
17	testing, in terms of looking at the
18	statistical performance of the model the
19	discriminate ability, the C statistic was
20	0.66, actually 0.67 in the validity data set
21	which is very good for this kind of a model.
22	Model calibration was also very

	Page 228
1	good. They looked at this using both
2	graphical techniques and other approaches as
3	well.
4	In terms of threats to validity 29
5	percent of the observations were missing data
6	on ejection fraction. Ejection fraction is
7	considered to be a very important clinical
8	risk factor for these types of models.
9	They imputed the missing data but
10	they used a very rough approach I guess for
11	imputation. And I quote, "We stratified by
12	gender and imputed the missing values to the
13	median of the corresponding groups."
14	So state of the art for imputation
15	is to use multiple imputation. I was a little
16	surprised that this approach was used. And I
17	was wondering if the measure developers could
18	maybe comment on this.
19	DR. CURTIS: Yes. I mentioned
20	we've been on this journey for 5 or 6 years so
21	sometimes it's a little hard to reconstruct
22	exactly what the logic was from that long ago.

	Page 229
1	Nevertheless there are two aspects
2	to, first, why EF is missing so frequently and
3	second, our approach to accounting for the
4	information that's conveyed by the missing
5	data.
6	So first off, the LVF is specified
7	to be an ejection fraction that an
8	information by the ejection fraction that is
9	available prior to the performance of the PCI.
10	So we obviously don't want to if a PCI goes
11	wrong and the patient has a large MI, has a
12	low EF we don't want to account for that in
13	the model. So it has to be LVF prior to the
14	PCI.
15	The patients who don't have
16	information about an ejection fraction before
17	the PCI typically are those that are being
18	done on an urgent or emergent basis. So, it
19	is very much colinear with patients who have
20	an ST elevation MI, patients with cardiogenic
21	shock or other highly morbid conditions. And
22	for that reason it's certainly not missing at

Page 230 1 random. To account for that, in addition 2 to imputation we actually also have a dummy 3 variable in the model for missing LVEF. 4 So that aggregates both as patients in whom it's 5 missing at random as well as those in whom it 6 is missing not at random, i.e., that they had 7 8 an emergent procedure. 9 But it does account for the 10 information that runs or is colinear with 11 missingness. And so we don't ignore the information that's conveyed by that. 12 13 In terms of the single versus multiple imputation, that's something I think 14 -- a decision that we made a long time ago 15 that was consistent with the inpatient 16 17 mortality model for ACC. And so for that reason I think we were trying to be harmonized 18 19 in terms of our approach. 20 It's something that we could re-21 look at or modify going forward. There's no reason that we couldn't do it from a 22

Page 231 1 statistical standpoint. So, in my mind this DR. GLANCE: 2 remains a bit of a threat to validity. 3 Some people would contest that if you're missing 4 this much data on a particular covariate that 5 it shouldn't even be included in the model. 6 And usually people typically, if 7 8 you're missing more than about 10 percent, maybe 20 percent of the data that's grounds 9 10 for sometimes not including it in the model. 11 The reference was made as to whether it's missing at random or missing not 12 13 at random. Without getting into a lot of the technical details what lots of folks will do 14 in this particular setting is they will create 15 a regression model based on all the other 16 17 available risk factors to predict the missing values for patients who have missing values. 18 If you can do this, if you can 19 predict a missing value using available 20 covariates then it is missing at random. 21 And I would suggest that probably in many of these 22

	Page 232
1	cases you could do that, that the data
2	actually is missing at random as opposed to
3	not missing at random. And I do think that is
4	a bit of a threat to validity.
5	Missing completely at random would
6	be if it was just well, I don't want to get
7	into too much of the details on this. But I
8	think it is an issue.
9	CO-CHAIR HALL: Jeptha, you
10	indicated that it was or did I
11	misunderstand that you said it was somewhat
12	colinear with emergency status.
13	DR. CURTIS: Right. I mean, it's
14	been a long time since we did those analyses,
15	but the patients in whom LVEF is not
16	available, it's not really necessarily missing
17	but it's not the test of ejection fraction
18	has not been performed, that is typically
19	those patients in whom you're under the gun to
20	perform a primary angioplasty.
21	So the STEMI patients for whom
22	we're trying to do a door to balloon time in

	Page 233
1	less than 90 minutes oftentimes we'll forego
2	doing an ejection fraction before the
3	procedure. So from that standpoint it is
4	colinear with the urgency or emergency
5	procedures.
6	And that it sort of fits in that
7	it's missing in about I'd say 10 to 20 percent
8	which is I'm sorry, I can't remember the
9	exact number. But that's the lion's share of
10	what's missing.
11	It's not necessarily missing the
12	information was available and just not
13	captured. I don't think that's what we're
14	looking at here.
15	DR. GLANCE: So by imputing it to
16	essentially a normal value which I believe is
17	what you're doing you are to some extent
18	disadvantaging hospitals that are taking care
19	of more emergencies compared to fewer
20	emergencies.
21	DR. CURTIS: Because we have a
22	dummy variable for missing information my

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	Page 234
1	understanding I said that before, yes.
2	We have a category for LVF of
3	and I miss the actual specifications, but it's
4	low EF, moderately low EF, normal EF, and
5	missing EF. I think there are four or five
6	categories for EF in the model. So missing
7	information is not ignored.
8	CO-CHAIR HALL: So the continuous
9	variable is moved to median and there's an
10	indicator as well. And we think it's probably
11	redundant information somewhat with at least
12	one other variable as well.
13	So, although it may not be perfect
14	it's probably less of a threat than it might
15	at first sound like it is.
16	DR. GLANCE: I will take back my
17	original comment. I did not realize, or I
18	didn't hear you when you said that you had an
19	indicator variable for missing. So I would no
20	longer qualify that as a significant threat to
21	validity.
22	CO-CHAIR HALL: Paul.

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	Page 235
1	DR. HEIDENREICH: I'd also say
2	that we have all this clinical data here that
3	we have in a couple of other measures. But
4	most of the measures you've been talking about
5	have claims data. And we're talking about
6	potentially even better risk prediction. So
7	I think in the big scheme of things any
8	threats to validity are probably pretty small.
9	CO-CHAIR HALL: I'm not seeing any
10	other cards raised so I would just like to
11	highlight if for my own understanding to make
12	sure.
13	The real significant updates to
14	the measure are that there were some changes
15	in the CathPCI registry variables which
16	warranted remodeling.
17	There's an improved strategy for
18	linkage, so that should again improve. I'm
19	just thinking in terms of overall validity.
20	The first issue, the update to Cath registry
21	data fields you would expect to improve the
22	validity overall as Paul just hinted at.

	Page 236
1	The improved linkage we would
2	expect to be an improvement overall.
3	The planned readmission algorithm
4	is an interesting aspect. You updated and
5	more or less expanded the planned readmission
6	algorithm such that at least in summary the
7	end result was that the overall crude planned
8	rate fell from 12.3 in the percent according
9	to the prior algorithm to 11.8 percent
10	according to the new algorithm.
11	In other words, you're giving more
12	institutions and providers more credit for
13	readmissions being planned. And so we're
14	minimizing hopefully we're minimizing or
15	reducing the chance of falsely throwing a flag
16	at them. Perhaps at some cost but that's what
17	we're doing.
18	The ICD crosswalk, not really an
19	issue yet.
20	And then updating your cohort code
21	again appears to be an improvement to the
22	overall validity of the measure.

	Page 237
1	Any other yes, Sherrie.
2	CO-CHAIR KAPLAN: So, you're not
3	opposed to doing multiple imputation if we
4	suggested that that might be an enhancement to
5	the missing data problem?
6	DR. CURTIS: I don't see it as
7	being a major barrier. I think that the vast
8	majority of the information that's conveyed by
9	missing LVEF is captured in that categorical
10	variable of missing EF. It is one of the more
11	powerful coefficients in the model. But yes,
12	we could certainly do that.
13	CO-CHAIR HALL: Well, I think as a
14	small ask maybe you could show in the future
15	what some of the modeling looks like with and
16	without in order to make the case of whether
17	it's really worth the effort.
18	Paul?
19	DR. HEIDENREICH: It just seems
20	like that would be an NQF sort of policy if
21	they want to make that, that one does multiple
22	imputation whenever you're doing imputation.

	Page 238
1	I would not recommend that but it
2	seemed like rather than have the different
3	groups suggest it here and there that should
4	be a standard policy.
5	CO-CHAIR HALL: I understand that
6	comment. I think NQF has shied away from
7	saying that models have to be performed in a
8	particular specific way as we've seen. We've
9	seen modeling done in a number of different
10	ways already.
11	So I understand your comment but I
12	would hazard a guess that it will remain the
13	judgment of the group as to whether a model is
14	appropriately specified. Larry?
15	DR. GLANCE: I think the issue
16	really isn't single imputation versus multiple
17	imputation. I think they're both fairly
18	straightforward to carry out. There's not a
19	huge difference in terms of doing that
20	mechanistically.
21	I think the issue is whether or
22	not you should use an indicator approach

	Page 239
1	versus an imputation-based approach. And I
2	think there's a lot of literature which will
3	show that estimates based on multiple
4	imputation are less biased that are based on
5	the indicator variable approach.
6	CO-CHAIR HALL: At the same time,
7	Larry, with all due respect that might be true
8	of the imputed variable itself. But in this
9	case, the variable probably is as part of a
10	large number of variables in this model the
11	information value may be redundant. And in
12	fact, there could be no value to multiply
13	imputing versus the current approach.
14	DR. GLANCE: So, agreed. We've
15	actually done some of that research and have
16	shown that there is a significant difference,
17	at least in the population that we looked at.
18	Whether or not that's generalizable to this
19	population I don't know. It's an empirical
20	question.
21	CO-CHAIR HALL: Agreed. We've
22	done the same thing in NSQIP so it's a common

	Page 240
1	topic that people spend money on beer on.
2	I think that, correct me if I'm
3	wrong, so far what we've stated is that the
4	current implementation seems to be a minimal
5	threat.
6	Other concerns or comments in this
7	category of validity? Not seeing any.
8	MS. SHAHAB: Voting for 2(b)
9	validity. One is high, two moderate, three
10	low, four insufficient and your time starts
11	now.
12	We have all the votes for 2(b)
13	validity. Two high, eighteen moderate, zero
14	low and zero insufficient.
15	CO-CHAIR HALL: Feasibility.
16	DR. GLANCE: So this measure is
17	based on a hybrid of clinical data and
18	administrative data. The administrative data
19	is just to identify which patients were
20	readmitted.
21	The clinical data is based on the
22	CathPCI registry. I would ask the measure

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	Page 241
1	developers to tell us what percentage of U.S.
2	hospitals that are currently performing PCIs
3	are in this particular registry.
4	MS. SLATTERY: So it's hard for us
5	to get to an exact number but based on what we
6	
7	CO-CHAIR KAPLAN: Could you state
8	your name, please?
9	MS. SLATTERY: Sorry. Lara
10	Slattery. Based on what we when we look at
11	our participating facilities against American
12	Hospital Association we estimate about 85
13	percent of current PCI hospitals are
14	participating in the registry, but that it
15	probably represents more about 90 percent of
16	the patients because it tends to be smaller
17	facilities, and usually smaller facilities
18	where they have a state reporting requirement
19	and no other incentive for them to join our
20	registry that are the facilities that are not
21	participating in.
22	DR. GLANCE: It's in line with the

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	Page 242
1	STS measure that we looked at previously.
2	CO-CHAIR HALL: I don't know
3	whether feasibility is the right category, but
4	could the developers comment on improvements
5	over time? This has been in play now for at
6	least 3 years, isn't that right?
7	DR. CURTIS: It's really only been
8	publicly reported last year. And so we don't
9	really have good information about
10	improvements over time.
11	And in fact, as we've made these
12	improvements to the model moving to direct
13	identifiers there's been enough changes that
14	I don't know if we have a good way of
15	surveillance as to what's been going on even
16	prior to public reporting. So it's probably
17	still a little bit early for us to comment.
18	CO-CHAIR HALL: Fair enough. But
19	there's a very clear implementation plan,
20	ongoing implementation. So it seems
21	acceptable from that perspective.
22	Other comments on feasibility from

	Page 243
1	other group members? I don't see any.
2	MS. SHAHAB: Voting for 3
3	feasibility. One is high, two moderate, three
4	low, four insufficient and your time starts
5	now.
6	We have all votes for feasibility.
7	Six high, thirteen moderate, one low and zero
8	insufficient.
9	CO-CHAIR HALL: Usability.
10	DR. GLANCE: So, I ask the measure
11	developers to provide us with information on
12	how this information is reported to
13	participating hospitals.
14	Do you report it both as a
15	continuous measure and also as a categorical
16	measure, meaning high-quality, low-quality and
17	average quality?
18	And if so in the most current
19	reporting period what proportion of the
20	hospitals were reported as being quality
21	outliers?
22	DR. CURTIS: So in the hospital-

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	Page 244
1	specific reports and actually in the public
2	reporting it's put both in buckets as well as
3	the overall adjusted risk-standardized rate.
4	So they receive that information.
5	In addition, in the hospital-
6	specific reports they receive information
7	about what hospital to what hospital the
8	patient was readmitted, the principal
9	diagnosis for that readmission and the time
10	frame of the dates of it.
11	So we're trying to encourage
12	through this hospital-specific report sort of
13	cross-fertilization and crosstalk across
14	hospitals so that they can try to work and
15	improve these rates.
16	In terms of the buckets, the
17	calculation of the outliers was done
18	completely in line with what's been done with
19	the other CMS measures. And I believe it was
20	about 2 percent of hospitals that were either
21	high or low outliers.
22	So relatively small number in part

	Page 245
1	because the data that we had available for the
2	measure with the Social Security numbers in it
3	that we needed to identify we only had 2 years
4	of data. So it was a little bit of probably
5	a challenge in terms of having enough
6	hospitals with enough volume that we could get
7	really good discrimination as to put them into
8	categories, but not that far out of line for
9	other publicly reported measures.
10	CO-CHAIR HALL: Was that 2 percent
11	at each tail, or 2 percent total?
12	DR. CURTIS: I think it was 2
13	percent total.
14	MS. SLATTERY: Lara Slattery. I
15	just wanted to clarify. Hospitals received
16	the feedback report by virtue of participating
17	in the registry regardless of whether they
18	opted in for the public reporting component.
19	So they all received the feedback reports.
20	And the hospitals for which we had
21	no data still received a benchmarking report
22	just for information purposes only.

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	Page 246
1	CO-CHAIR HALL: Additional
2	comments or concerns on usability? Karen?
3	DR. JOYNT: I know someone just
4	said this, but is there any way to or
5	should we be thinking about ways to update the
6	measure such that more than 2 percent can be
7	identified as something? Or do you feel that
8	the value is not about the identification as
9	an outlier but about the benchmarking?
10	DR. CURTIS: I think we've had
11	this discussion in different forms all
12	morning.
13	DR. JOYNT: Do you know what
14	happened to those that were identified? Like
15	can you look over the last 3 years and see
16	what prior to the reporting or identification
17	as outliers, sort of how your measure helps
18	people move? Do you have the information on
19	that?
20	DR. CURTIS: Again, we've only had
21	one year of public reporting which is last
22	March. And so we have not seen what's

	Page 247
1	happened over time.
2	And I think your work and others
3	has questioned whether or not it does make a
4	dramatic change as opposed to incremental
5	change.
6	And I think the jury's still out
7	on how any of these measures can be used to
8	change national performance. I think we are
9	seeing encouraging trends but nothing definite
10	yet.
11	But I think of it as an
12	implementation question as to where do you
13	draw the buckets and where do you draw the
14	line. Are you very restrictive or are you
15	more permissive in terms of categorizing
16	hospitals as better than or worse than.
17	And again, we in this case tried
18	to be as consistent as possible with other CMS
19	measures.
20	CO-CHAIR HALL: Could the
21	developers comment briefly on the unintended
22	negative consequences remarks in the measure

	Page 248
1	materials?
2	DR. CURTIS: I think as we've
3	discussed for other measures there's always
4	the possibility of unintended consequences
5	where you could have at worst case aversion on
6	the basis of readmission rates. And that's
7	really nothing that you can necessarily
8	prophylaxe against.
9	I think it's probably lower stakes
10	for readmissions in procedures than it is
11	perhaps for mortality. Just that would be my
12	initial impression. But I think it's
13	something that has to be monitored and can be
14	monitored.
15	And we've looked at for other,
16	specifically for mortality we've been looking
17	at whether or not expected case mix and
18	predicted risk has changed over time in states
19	that publicly report PCI mortality and have
20	not seen dramatic changes in sort of I'm
21	sorry, not the predicted but the expected
22	mortality at the states that do have public

Page 249 1 reporting. CO-CHAIR HALL: Other thoughts or 2 3 concerns on usability? Seeing none. MS. SHAHAB: Voting for usability 4 and use. One high, two moderate, three low, 5 four insufficient information and the time 6 Three more votes. 7 starts now. 8 CO-CHAIR HALL: Can everybody do it one more time? Sorry. 9 10 MS. SHAHAB: We have all the votes 11 for usability and use. Three high, fourteen moderate, three low and zero insufficient 12 13 information. CO-CHAIR HALL: Any comments in 14 summary before an overall vote? I see Karen. 15 16 DR. JOYNT: I just have another 17 clarifying question. You can tell I've not been on this committee before. 18 This came up with another measure 19 20 yesterday that we were re-approving or whatever. Is the burden of proof that it has 21 been implemented, or that it has made X amount 22

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	Page 250
1	of change? Or what is the sort of what is
2	our responsibility in terms of looking at its
3	longitudinal performance in continuing it
4	forward?
5	CO-CHAIR HALL: I'll give my
6	opinion on that but I'll be corrected if I'm
7	wrong.
8	I think when there's a clear plan
9	for ongoing implementation that's kind of set
10	one. And if that plan has been and is being
11	carried out in good faith but may not have
12	results yet I think that should meet our
13	adequacy threshold. That's my opinion.
14	DR. BURSTIN: Just to add to that,
15	last year Karen had to leave, but she led
16	an effort with a task force to update our use
17	and usability criterion.
18	It used to just be usability which
19	basically was it in use. And I think what we
20	really heard from the community, that's not
21	enough.
22	So in use is really important.

	Page 251
1	And by the 3-year window we want to see that
2	it's being used in an accountability
3	application within two of those cycles. We
4	want to see evidence of public reporting.
5	But I think increasingly what
6	we're trying to see is in addition to that,
7	and this is not a must-pass like evidence and
8	scientific acceptability, but we want to in
9	fact be able to see over time that measures in
10	use have helped to move the needle hopefully
11	positively.
12	We also want to be cautious as
13	we've talked a lot about over the last couple
14	of days that we also haven't seen any
15	unintended consequences as a result of that
16	use.
17	So it's something we'd love your
18	input on. This is really just I think the
19	first year that we've actually implemented the
20	broader lens on use and usability. Bruce got
21	it right.
22	DR. CURTIS: Can I just follow up?

	Page 252
1	I mean, I think there are different ways that
2	you can measure impact.
3	I think when we developed this
4	measure nobody was talking or thinking about
5	readmissions after PCI. And certainly the
6	whole healthcare system has really evolved in
7	their consideration of the importance of this
8	particular aspect for pros and cons and
9	differences of opinions. It's out there.
10	I can say that over the past 5
11	years since we published our first paper in
12	JACC just describing the readmission rates
13	there has been an abundance of literature
14	coming out examining the issue.
15	So I think we have moved the
16	conversation on this. The next will be to see
17	can we move the actual rates.
18	CO-CHAIR HALL: The notion that
19	you're only calling out going back to one
20	of Karen's earlier remarks, that you're only
21	calling out a percent or so at each tail. Do
22	you have any internal plans or deliberations
	Page 253
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1	around changing that as potentially a way to
2	drive change faster?
3	DR. CURTIS: So let me throw that
4	out. I think we are going to be cautious as
5	an organization in terms of how we call out
6	hospitals.
7	And I think that the real value
8	for me is in reporting this information back
9	to hospitals, that they get their risk-
10	standardized rates, that they have that
11	they're valid, that they're believable and
12	that they are usable.
13	And the usable is I think what
14	needs to evolve most rapidly. I have spent
15	the last four years of my life as opposed to
16	sorry, in addition to working on this
17	measure working on developing the evidence
18	that will support reducing these readmission
19	rates.
20	And we're in the last stages of a
21	mixed method study understanding both the
22	qualitative and quantitative strategies

Page 254 1 associated with lower readmission rates. So I think what the ACC and what 2 we're talking about, and I'm not going to 3 guarantee that this will be done because I 4 don't speak for them in that regard, but we're 5 trying to create an environment where we will 6 take that evidence, combine it with this 7 information and create a campaign or an effort 8 9 to try and systematically reduce these rates 10 at hospitals. 11 And I think much more so than identifying high and low outliers that's the 12 13 way we're going to push things forward. You have to develop the evidence. 14 You have to package it up in a toolkit or some other 15 16 change -- some process that promotes change in 17 a positive force much more so than putting it in buckets. 18 MS. SLATTERY: And so if I can 19 just add to Dr. Curtis' remarks which are 20 totally in line and I agree with. 21 This is our first foray out with a 22

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	Page 255
1	measure that the hospitals were offered the
2	opportunity to publicly report on.
3	The College has a larger portfolio
4	within PCI but also this year we will be
5	implementing it in the ICD implantable
6	cardiodefibrillator. We do have plans to
7	expand public reporting opportunities for all
8	hospitals across our registry initiatives.
9	So, once we start to look at that
10	contextually we want to ensure consistency in
11	how we report the information out. So we
12	actually it's a separate workgroup that is
13	taking that on and will be working through
14	those approaches. And we will likely evolve
15	it over time.
16	But I agree, it's meant to drive
17	quality improvement and be engaging a dialogue
18	and useful for consumers. So it's hard to
19	balance out all of those.
20	CO-CHAIR HALL: Thank you. Larry?
21	DR. GLANCE: I'd just like to add
22	my thoughts on this discussion.

1	
	Page 256
1	I think that the burden on showing
2	that there is an improvement in population
3	outcomes with a quality measure should not
4	rest primarily with the measure developer.
5	I think that as long as the
6	quality of the information that's being
7	provided in a quality metric is deemed to be
8	acceptable or high then it really is up to the
9	end user to make those to use that
10	information and improve population outcomes.
11	So I think we should be hesitant
12	about looking for improvement in population
13	outcomes as a determining factor in whether or
14	not a quality metric is usable or not usable.
15	CO-CHAIR HALL: Cristie.
16	MS. TRAVIS: Just a clarifying
17	question on the public reporting because you
18	said it just a couple of times. Is this a
19	voluntary public reporting on behalf of the
20	hospitals? And if so, about what percentage
21	of those who are reporting are actually being
22	willing to be voluntarily reported publicly?

	Page 257
1	DR. CURTIS: That's a great
2	question. Yes, it is a completely voluntary
3	public reporting. And if hospitals opted out
4	there was no indication that they had opted
5	out. So it was simply those hospitals that
6	opted in, at least on Hospital Compare.
7	Of the I think 1,200 hospitals
8	that had met our reporting thresholds 350
9	roughly decided to participate. I will say
10	personally that was about 320 more than I
11	thought were going to be willing to do it. So
12	I thought for an initial year's effort for
13	voluntary public reporting it was very
14	successful.
15	MS. SLATTERY: I'll also just
16	since we talked a little bit about
17	implementation and you want to understand
18	where we're going with it.
19	I think that that's incredibly
20	admirable, the hospitals that opted in.
21	Because one of the things for you to
22	understand is we went from providing the

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	Page 258
1	reports to the hospitals to giving them a
2	window of about 6 to 8 weeks to make a
3	decision to voluntarily report.
4	The lion's share of those 300 and
5	some odd hospitals opted in at that point.
6	There was only one other opportunity for those
7	hospitals to opt in to have it publicly
8	reported and that's why the number is stable.
9	Dr. Curtis also referenced the
10	fact that if you chose not to report there's
11	nothing reflected on Hospital Compare.
12	Moving forward the ACC has made
13	the decision that we will be making it more
14	obvious to consumers if a hospital is
15	participating in our registries and had the
16	opportunity to be able to report that and
17	elected not to. And then differentiating some
18	of the categories of decision-making in
19	displaying that information out. So moving
20	forward there will be more.
21	And moving forward we will
22	continue to partner with CMS on reporting this

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	Page 259
1	data on Hospital Compare because we think that
2	that is a very valuable resource for
3	consumers.
4	We also will have a complementary
5	effort within our organization. And the
6	reason for doing that is it allows for a
7	mechanism for more rapid dissemination of
8	information.
9	So, right now data gets reported
10	to Hospital Compare on a quarterly basis. The
11	infrastructure we will put in place on our
12	website will allow that as soon as a hospital
13	makes the decision to publicly report it will
14	be made available to the public. And we can
15	control the infrastructure better to be able
16	to do that.
17	CO-CHAIR HALL: Do you know if the
18	performance distribution in the reporting
19	group is different than in the overall group?
20	DR. CURTIS: Yes.
21	CO-CHAIR HALL: Is there a concern
22	that the public is getting the wrong

Page 260 1 impression? (Laughter) 2 So we did have some 3 MS. SLATTERY: of the low-performing hospitals opt in for 4 voluntarily publicly reporting. 5 I mean, again, I do think that in 6 fairness to them the window of opportunity for 7 which to make a decision was incredibly tight. 8 9 10 So, even for the second sweep they 11 at best had about 12 weeks to make a decision, get leadership buy-in within their 12 13 organization, get legal counsel to review it, and get the paperwork back to us for it to 14 15 appear. 16 So we think that moving forward 17 over time it will be easier for hospitals to get on board with this. 18 CO-CHAIR HALL: Thank you. 19 Sherrie? 20 CO-CHAIR KAPLAN: To follow on to 21 Larry's comment. And that is sensitivity to 22

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	Page 261
1	change is one of these awkward things where
2	you don't know what actually manipulations
3	were going on that your measure should be
4	responsive to. So, the attribution to
5	sensitivity to change implies efforts to
6	quality improve where you don't know what
7	those are. And especially at the tails where
8	we are the worst at estimating where a
9	hospital actually might be.
10	Regression to mean pops to mind.
11	And you think the trouble around the tails
12	without actually having a response to chase.
13	So, is there is there any
14	effort and distributional scoring will
15	always have this property. Somebody always
16	loses. So is there any push towards trying to
17	find the mutable point beyond which a
18	threshold one would declare a hospital as
19	either better or worse or whatever?
20	Or is there any effort along those
21	lines to shift away from distributional
22	scoring?

	Page 262
1	DR. CURTIS: I think it's a great
2	question and something that the field as a
3	whole I think continues to struggle with. And
4	I think we always default to the
5	distributional because it's comfortable and
6	familiar.
7	I think for readmissions
8	specifically we don't know what the floor is.
9	And I think until we kind of know and they
10	start bunching up on the lower side I think
11	it's reasonable to use the distributional but
12	be attuned to the fact that since we know that
13	there are unplanned readmissions that are not
14	preventable that the goal is not to go to
15	zero.
16	CO-CHAIR HALL: Larry?
17	DR. GLANCE: One last comment.
18	Since our discussion now is about endorsement
19	I'd like to point out that this like many of
20	the other measures that we have looked at over
21	the last two days is I believe a very robust
22	measure methodologically speaking.

	Page 263
1	What differentiates this measure
2	from many of the other measures that we've
3	looked at is that this particular measure is
4	largely based on clinical as opposed to
5	administrative data.
6	And I think that's really a very,
7	very important qualifier because clinical data
8	is believed by most to be much more accurate
9	and therefore have much greater face validity
10	compared to administrative data.
11	CO-CHAIR HALL: Thank you. Same
12	comments we had about the STS discussion
13	earlier on. Jeptha?
14	DR. CURTIS: Not to shoot chart-
15	based measures in the foot, but I do think
16	it's important to recognize that they have
17	different strengths.
18	And I think that, yes, we're
19	really good at knowing whether or not a
20	patient has diabetes or what the creatinine
21	was and things like that. And that is
22	important in terms of risk adjustment.

	Page 264
1	We're not as good as an
2	administrative model I don't think at
3	measuring frailty and accounting for frailty
4	in a robust fashion. So, I think it's I
5	appreciate the support and I hope that it's
6	taken into account but I think there are two
7	different schools of thought and there are
8	competing strengths.
9	CO-CHAIR HALL: Thank you. I see
10	no cards up so let's move to vote overall.
11	MS. SHAHAB: Voting for overall
12	suitability for endorsement. One yes, two no
13	and time starts now. Two more votes. One
14	more.
15	CO-CHAIR HALL: There's some empty
16	chairs down at the end of the table.
17	MS. SHAHAB: All votes are in.
18	For overall suitability for endorsement for
19	measure 0695 Hospital 30-day Risk-standardized
20	Readmission Rates Following Percutaneous
21	Coronary Intervention, 20 yes, zero no.
22	CO-CHAIR HALL: We thank the

Page 265 1 developers and we'll ask the Yale CMS teams to come back to the table for the next measure. 2 3 CO-CHAIR KAPLAN: Okay, welcome to the developer. I'd ask you to do what we've 4 been doing which is briefly introduce 5 6 vourself. 7 And also as soon as you speak 8 remind you please say again your names because 9 the recorder back in the corner can't see you. 10 And then briefly introduce your measure. DR. BERNHEIM: 11 Hi, this is Susannah Bernheim. I'm a director of quality 12 13 measurement for the Yale Core Team many of 14 whom you have met today. Nihar, do you want to introduce 15 16 yourself? 17 DR. DESAI: My name's Nihar Desai. I'm a cardiologist at Yale and an investigator 18 at the Center for Outcomes, Research and 19 20 Evaluation. 21 MR. AMIN: Do you have anyone on the phone? 22

	Page 266
1	DR. BERNHEIM: I don't know that
2	we have anyone on the phone. Do we have
3	anyone on the phone?
4	(Laughter)
5	DR. BERNHEIM: We have some
6	support in the back and CMS here.
7	MR. AMIN: Okay.
8	DR. BERNHEIM: But I don't believe
9	that there's anybody on the phone for this
10	measure.
11	Okay, so I will just say a couple
12	of words about this measure. This is a 30-day
13	all-cause unplanned readmissions for
14	hospitalizations following hospitalizations
15	with an acute myocardial infarction.
16	This measure originally came to
17	NQF in 2008 and this is its first time back
18	for full re-endorsement which is a little
19	longer than usual just because of the cycles
20	of projects.
21	It has been in public reporting
22	through the inpatient quality reporting

	Page 267
1	program since 2009 and last year was included
2	in the first year of the hospital readmission
3	reduction program.
4	It is designed much like our other
5	measures where you look at a cohort of AMI
6	patients. We look at readmissions 30 days
7	later. I will talk a little bit about the
8	planned readmission algorithm in a moment to
9	identify unplanned readmissions. It uses the
10	same hierarchical modeling approach as our
11	other measures and it is a claims-based
12	measure.
13	There have been a number of
14	changes over the years that are detailed in
15	your in the application. I'll just
16	highlight the important ones.
17	When it first was reported we
18	moved from a one-year measure to a three-year
19	measure because many hospitals do not have a
20	huge volume of AMI cases and that allowed us
21	to report on a greater number of hospitals.
22	Other key changes. Early on we

	Page 268
1	excluded patients who were discharged against
2	medical advice. That's true now with all of
3	the readmission measures.
4	In reporting I believe two years
5	ago this measure was expanded to include data
6	from VA hospitals. So in the publicly
7	reported measure it's now all CMS hospitals
8	and patients hospitalized originally with
9	their AMI at a VA hospital. So that was a
10	neat collaboration between CMS and the VA
11	which took a fair amount of work.
12	And then the one other big change
13	has been that as part of the development of
14	our hospital-wide readmission measure which
15	this committee is going to be talking about
16	later we created an algorithm that used claims
17	codes to try to define readmissions that were
18	planned or likely to be scheduled in advance,
19	largely procedural readmissions that were not
20	associated with an acute diagnosis code. And
21	there's a long algorithm that does that.
22	And we developed that first in a

	Page 269
1	hospital-wide cohort and then we carefully
2	checked it against the condition-specific
3	cohorts.
4	And last year NQF held an ad hoc
5	review of our condition-specific measures
6	including this measure just to look more
7	closely at this planned readmission algorithm.
8	So that piece of this measure has come before
9	committee at NQF previously, but the whole
10	measure hadn't come back at that point for re-
11	endorsement.
12	I think an interesting case was
13	made earlier that improvement doesn't prove
14	that the measure works or not. But I will say
15	because we're really excited about it that in
16	the 3-year cycle that was reported last
17	December we are seeing for the first time
18	declining national AMI readmission rates. And
19	that's in the context of also decreasing
20	admissions for AMI and potentially higher
21	severity admissions. And big declines in
22	mortality.

	Page 270
1	And so I think we're very
2	reassured around questions of unintended
3	consequences that the decline in readmissions
4	is happening in the context of other
5	improvements around AMI.
6	I think that that's probably
7	enough of a quick overview of who know this
8	measure well. But obviously we'll answer
9	questions.
10	CO-CHAIR KAPLAN: Thank you very
11	much. Paul, do you want to go first on the
12	evidence?
13	DR. HEIDENREICH: Yes. I think
14	there is we don't necessarily have a lot of
15	evidence to know exactly how hospitals are
16	improving. Clearly in 2012 there was a sudden
17	drop in readmissions for MI as well as all.
18	It seems to be most hospitalizations for
19	Medicare patients. So I think there's clearly
20	a strong rationale that one could make
21	improvements.
22	CO-CHAIR KAPLAN: Larry? Nothing

	Page 271
1	to add? Other comments? Ready to vote?
2	MS. SHAHAB: Voting for 1(a)
3	evidence. One is yes, two is no and your time
4	begins now.
5	All votes are in for 1(a)
6	evidence. Nineteen yes, zero no.
7	CO-CHAIR KAPLAN: Great.
8	Performance gap. Paul?
9	DR. HEIDENREICH: So that has
10	narrowed but I think still remains important.
11	I think the 10 percent/90 percent went from
12	17.9 to 19.4 several years ago and it looks
13	like with the last drop was down to 17.3 and
14	18.3. So the higher end clearly dropped
15	although still a reasonable difference between
16	the groups. And the overall rate some might
17	argue is still too high. So I'd say there is
18	a significant evidence performance gap.
19	CO-CHAIR KAPLAN: Larry? Other
20	comments? Ready to vote?
21	MS. SHAHAB: Voting for 1(b)
22	performance gap. One is high, two moderate,

Page 272 1 three low, four insufficient. And the time begins now. Just one more vote. 2 All votes are in for 1(b) 3 performance gap. Nine high, ten moderate, 4 zero low and zero insufficient. 5 6 CO-CHAIR KAPLAN: Thank you very much. Priority? 7 DR. HEIDENREICH: It has been, I 8 9 think probably still remains a priority for 10 the government, for CMS to improve readmission 11 rates for MI. CO-CHAIR KAPLAN: Larry? Other 12 13 comments? Ready to vote? MS. SHAHAB: Voting for 1(c) high 14 priority. One is high, two moderate, three 15 low, four insufficient. And the time begins 16 17 now. All votes are in for 1(c) high 18 priority. Fourteen high, five moderate, zero 19 low and zero insufficient. 20 21 CO-CHAIR KAPLAN: Thank you. Scientific acceptability, reliability. Paul. 22

	Page 273
1	DR. HEIDENREICH: So, I think this
2	has been felt to be moderate. I think they
3	used a test/retest with a random sample and I
4	see a reported ICC of 0.38 which is, you know,
5	I think relatively common for this type of
6	data.
7	CO-CHAIR KAPLAN: Larry? Other
8	comments? Ready to vote? Oh, you have a
9	comment.
10	DR. BERNHEIM: I just wanted to
11	clarify it was brought up in the CABG just
12	because people had taken some interest in
13	this. We did do some work because when we
14	create these ICCs we're not using six years of
15	data to get a three-year sample size. So Lisa
16	explained earlier we put together a correction
17	factor. And we can share those details.
18	And when you do that for this
19	measure it goes up to 0.48 just so people
20	know. It's stronger. Our best estimate of
21	what it would be with a full three-year sample
22	size.

1	
	Page 274
1	CO-CHAIR KAPLAN: Certainly well
2	within what we've been seeing. So it's not
3	exactly like this is a huge departure but
4	thank you for that clarification.
5	Other comments? Ready to vote?
6	MS. SHAHAB: Voting for 2(a)
7	reliability. One is high, two moderate, three
8	low and four insufficient. And the time
9	starts now.
10	All votes are in for 2(a)
11	reliability. Three voted high, sixteen
12	moderate, zero low and zero insufficient.
13	CO-CHAIR KAPLAN: Thank you.
14	Validity, Paul?
15	DR. HEIDENREICH: Yes, there's
16	been I think with the original submission
17	I'm not sure if things were updated but the
18	model, the overall model's discrimination had
19	C statistics close to 0.6 and slightly under.
20	It sounds like interestingly it
21	was I think when it was tested in the CCP
22	project which had actual chart review it

	Page 275
1	sounded like there was a very similar C
2	statistic. So it didn't seem like you were
3	losing a whole lot from using, or if any at
4	all of using administrative data.
5	So that seems to be I think very
6	reasonable for this type of data. I think the
7	exclusions as described are reasonable. I
8	don't think there's been if anything the
9	slight changes over time have improved
10	improved the model since it was last approved.
11	I didn't feel there were any
12	significant issues with missing data but we'll
13	see if anyone else has concerns.
14	CO-CHAIR KAPLAN: Larry? Can you
15	clarify the C statistic for us? Because I
16	want to make sure everybody understands. What
17	the magnitude of it.
18	DR. BERNHEIM: Sure. So you were
19	right. When it was first developed, the
20	technical report from development I think it
21	was 0.58. In the most recent year of data
22	we look at it each time it gets run. The most

	Page 276
1	recent three years it was 0.64.
2	DR. HEIDENREICH: Oh, I actually
3	remember I did have one question. The I
4	know it's been then tested in California data
5	so now you can have an 18 and over measure.
6	But I didn't it wasn't clear to me if there
7	was a significant improvement or decrease in
8	the model's performance.
9	DR. BERNHEIM: So that's a great
10	question that I don't remember the answer to.
11	In general our measures often do slightly
12	better in the all-payer data sets. We think
13	that's because the comorbidities are even more
14	powerful predictors in younger populations
15	that have fewer of them.
16	And I can quickly find you the
17	answer to this for this particular one. So
18	0.67 was the discrimination for the 18 and
19	over measure. And the correlation between the
20	two was 0.998.
21	CO-CHAIR KAPLAN: Larry?
22	DR. GLANCE: So, just as a quick

	Page 277
1	point of clarification. As was alluded it's
2	very common to see a model perform differently
3	in data sets.
4	And specifically if you're going
5	to use an all-payer data set you expect to see
6	more heterogeneity in the patient population
7	and therefore as a result of that you'll see
8	better discriminatory power.
9	CO-CHAIR KAPLAN: Thank you.
10	Other comments? Frank.
11	DR. BRIGGS: What was the impact
12	of expanding the planned readmission
13	algorithm?
14	DR. BERNHEIM: Great question.
15	Again, numbers I don't have at the tip of my
16	tongue.
17	It reduces the overall readmission
18	rate very slightly. Although for this
19	measure, let's see if I can find it rapidly
20	for you.
21	CO-CHAIR KAPLAN: So refreshing to
22	see people throwing pages as opposed to

	Do
	Page 278
1	scroll.
2	(Laughter)
3	DR. BERNHEIM: Exactly, not be on
4	a screen. Okay. Planned readmission
5	algorithm I don't see here.
6	If somebody on the phone from our
7	team has this handy please speak up. And I
8	will otherwise find it quickly but it'll take
9	me just a minute.
10	Can the operator open the lines
11	and make sure that somebody on our team is
12	available to answer this question quickly?
13	Because they have it at their fingertips
14	quicker than I do.
15	MS. KHAN: What are their names?
16	The people that are on the phone.
17	DR. BERNHEIM: Chanch Nabat, are
18	you there?
19	OPERATOR: All lines are open.
20	DR. BERNHEIM: Okay, great.
21	Anybody on the Yale team have this number
22	handy quickly so I don't have to make this

Page 279 1 poor, tired committee wait while I flip pages? MS. GEARY: Hi, this is Lori Geary 2 3 at Yale. Can you hear me? DR. BERNHEIM: Yes. 4 MS. GEARY: We are pulling that up 5 now. Bear with us one minute. 6 The easiest place 7 DR. BERNHEIM: 8 may be the NQF application where we brought it 9 back last year. 10 MS. GEARY: Okay. From version 1 11 to version 2 it went from 19.7 to 19.0. I'm sorry, 18.7. 12 13 DR. BERNHEIM: Okay, so a percentage point. We should have gone with my 14 guess, I was right. Thank you, Lori. 15 16 MS. GEARY: Okay. 17 CO-CHAIR KAPLAN: Thanks for that 18 clarification. Any other comments? Ready to vote? 19 20 MS. SHAHAB: Voting for 2(b) 21 validity. One is high, two moderate, three low, four insufficient and your time begins 22

Page 280 1 now. All votes are in for 2(b) 2 3 validity. Four high, fifteen moderate, zero low and zero insufficient. 4 CO-CHAIR KAPLAN: Thank you. 5 Feasibility? 6 DR. HEIDENREICH: Based on claims 7 8 data, highly feasible. 9 CO-CHAIR KAPLAN: Larry? Other 10 comments? Ready to vote? 11 MS. SHAHAB: Voting for feasibility. One high, two moderate, three 12 13 low, four insufficient. Time starts now. All votes are in for feasibility. 14 Eighteen high, one moderate, zero low, zero 15 insufficient. 16 17 CO-CHAIR KAPLAN: Thank you. 18 Usability and use. Paul? DR. HEIDENREICH: Well, it already 19 20 is being used. And you know, one could argue it's been successful given that it's been used 21 both I think for public reporting as well as 22

Page 281
for payment.
And we've seen the expected
changes at least within 2012.
CO-CHAIR KAPLAN: I have one quick
question. If it were possible, would it be
possible to investigate actually the
sensitivity to places where you actually knew
there were efforts underway to improve this?
And so responsiveness to change could actually
be estimated.
DR. BERNHEIM: So I think what
you're asking is could we focus in on places
that we know are making a big effort around
this and show that those efforts are playing
out. I mean I think that's a great research
question. It's not something we've done on
our team, but I think it is an important link
that the research is slowly building to show.
And there are some trials out there that have
shown particular interventions work AMI
patients.
CO-CHAIR KAPLAN: Thank you. That

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	Page 282
1	raises the specter of the R word so we're
2	moving on really quickly. So, for usability
3	and use are there any other sorry, Karen?
4	DR. JOYNT: I think this is a
5	great example of ways in which the same
6	measure can be used a lot of different ways.
7	In the way that it's used in
8	public reporting the of the 4,464 hospitals
9	in the U.S. 23 are identified as being better
10	than average, 2,327 are no different, 29 are
11	worse and 2,085 are number of cases too small.
12	That's a lot of effort for
13	hospitals to make for this amount of
14	discrimination. And I know they get more
15	information than this.
16	On the completely opposite side as
17	the weight of the readmissions penalty is
18	calculated in which there's no uncertainty
19	built into the model if you're 0.001 percent
20	worse than predicted on dollars, not even on
21	rates, that you will in theory get a penalty.
22	And I don't know that there's any

	Page 283
1	way in this committee to address the different
2	ways in which something can be used, but it
3	certainly points out that the statistical
4	arguments that we have about the way that this
5	thing is done can go any number of different
6	ways when things are put forward.
7	And to me the way that things are
8	used actually is really important to how the
9	measure is going to work. And I might
10	personally choose different ways of
11	calculating this based on whether it was going
12	to be used for public reporting or for pay-
13	for-performance or whatever this is going to
14	be.
15	So I just have concerns about this
16	sort of blanket blessing of measures when they
17	can be used in such vastly different ways.
18	That may be a bigger problem in this
19	particular measure but I think I'd be
20	interested in hearing from the developers sort
21	of how we should think about what your model
22	can do.

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	Page 284
1	The data to suggest exactly what
2	you're saying which is what we know can and
3	can't work to improve readmissions is not
4	great. And I'd be interested in knowing from
5	your work are the people that are identified
6	as outliers doing something differently.
7	What's happening as a result of
8	the way that this metric is being used?
9	CO-CHAIR KAPLAN: Hold on a
10	second.
11	CO-CHAIR HALL: A couple of
12	things. Karen, your sentiment is a
13	longstanding one. It's been heard in many,
14	many NQF forums over the years. And in fact
15	I think I won't jump the gun, but in the
16	upcoming white paper coming out from NQF there
17	may be some commentary about NQF increasing
18	its guidance around recommended uses or uses
19	where a particular measure seems most
20	appropriate. So we won't change that aspect
21	of it today but your sentiment has been heard
22	many, many times.

	Page 285
1	If there's
2	DR. JOYNT: And if this needs to
3	be tabled I'm totally fine with that. I'm
4	struggling a little bit with sort of the
5	questions around usability for us to sort of
6	say as it's being used what are the negative
7	consequences, what's happened. And we don't
8	really get that data
9	CO-CHAIR HALL: Absolutely. And
10	that's been I think we all sympathize.
11	That's absolutely been the case.
12	But so if there is something you
13	would like our developers to state please
14	rephrase that. I didn't want to cut you off,
15	but I do want you to know that that is a
16	longstanding concern and there may be some
17	change in the air around that concern. But it
18	won't happen today.
19	Is there anything you do want the
20	developers to state? Okay. Any other
21	comments or concerns then?
22	CO-CHAIR KAPLAN: Just to say that

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	Page 286
1	some of us share your pain. And the hopefully
2	MAP group will begin to address these kinds of
3	issues.
4	So there's a safety clause built
5	in but it's not sufficient for some of us to
6	kind of feel cozy about these decisions.
7	Okay, Susannah.
8	DR. BERNHEIM: I will decline from
9	commenting on the pieces that you guys want
10	not commented on. I just wanted to make sure
11	that people know that a part of the work that
12	we do with CMS is explicitly monitoring for
13	unintended consequences. And we do some
14	surveillance work. And there is a constant
15	measure or maintenance process. So just in
16	terms of the unintended consequences piece
17	that is a part of the expectation as part of
18	the measure life cycle.
19	CO-CHAIR KAPLAN: Thanks for that
20	clarification. Any other comments? Are we
21	ready to vote usability and use? Okay.
22	MS. SHAHAB: Voting for usability

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	Page 287
1	and use. One high, two moderate, three low,
2	four insufficient information and the time
3	starts now.
4	All votes are in for usability and
5	use. Four high, fourteen moderate, zero low
6	and one insufficient information.
7	CO-CHAIR KAPLAN: Thank you very
8	much. Moving onto suitability for
9	endorsement. Paul?
10	DR. HEIDENREICH: I think no
11	additional comments. I'd say it meets
12	endorsement in my opinion.
13	CO-CHAIR KAPLAN: Larry?
14	DR. GLANCE: I think this measure
15	is also very typical of most of the CMS
16	measures that we've heard in other measures.
17	It's very robust in terms of the methodology
18	and I would also vote for endorsement.
19	CO-CHAIR KAPLAN: This is
20	Washington so I won't say "robustitude" in
21	this audience but it does so we are voting
22	on its suitability for endorsement. Ready to

	Page 288
1	vote?
2	MS. SHAHAB: Voting on overall
3	suitability for endorsement. One yes, two no.
4	Time starts now.
5	All votes are in for overall
6	suitability for endorsement for measure 0505
7	Hospital 30-day All-cause Risk-standardized
8	Readmission Rate Following Acute Myocardial
9	Infarction Hospitalization. The results are
10	17 yes, 2 no.
11	DR. BERNHEIM: Thank you.
12	CO-CHAIR KAPLAN: Thank you to the
13	Yale group.
14	CO-CHAIR HALL: We'll invite the
15	next Yale group to the table.
16	(Laughter)
17	CO-CHAIR HALL: We'll be
18	discussing 2539 Seven-day Risk-standardized
19	Hospital Visit Rate after Outpatient
20	Colonoscopy. I think in some sense there's a
21	little bit of a twist compared to some other
22	topics we've discussed. So we'll wait for our
Page 289 1 developers. No, that was not a pun. 2 3 (Laughter) CO-CHAIR HALL: So we have 18 4 people in the room. No one's permitted to 5 leave because we lose our quorum. 6 So raise your hand if you have to do a number one. 7 8 (Laughter) 9 MS. KHAN: We should be done by 2 10 I think. 11 CO-CHAIR HALL: We'll continue to push on. Our colleagues from Yale, please 12 13 introduce yourselves and your measure. DR. DRYE: Hi, I'm Elizabeth Drye 14 15 from Yale. 16 DR. RANASINGHE: My name's Isuru 17 Ranasinghe from Yale. DR. DRYE: I was just going to say 18 following up on Bruce's point that this is a 19 little bit of a twist. This is the first 20 21 measure we're bringing to NQF that is for outcome of ambulatory care. So we're really 22

	Page 290
1	excited about it. It is new terrain. And
2	Isuru led the work and he's going to walk us
3	through.
4	DR. RANASINGHE: Okay. So it's
5	Isuru here again.
6	I'll start off by doing a quick
7	summary of the measure and the rationale for
8	the measure.
9	So the measure is a measure of
10	unplanned hospital visits following outpatient
11	colonoscopy. And that is colonoscopies
12	performed in hospital outpatient department
13	ambulatory pre-surgical centers and physician
14	office settings.
15	The denominator for this measure
16	is low or moderate risk colonoscopy
17	procedures. And based on our inclusion and
18	exclusion criteria we actually capture about
19	94 percent of all outpatient colonoscopies
20	performed.
21	The numerator for this measure is
22	unplanned hospital visits within 7 days of the

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	Page 291
1	procedure. And hospital visits include ED
2	admissions, observation stays and inpatient
3	admissions.
4	Now, that's a very broad patient-
5	centered outcome that captures adverse events
6	that are related to the bowel prep, the
7	anesthesia itself and the procedure.
8	Now, this measure is really
9	important because of four critical reasons and
10	I'll outline them.
11	First is that colonoscopy is
12	incredibly common. So this is the most common
13	procedure performed in the outpatient setting.
14	We see an outcome rate of about 16.2 per 1,000
15	procedures in the Medicare data and we see
16	significant variation between 8 to 20 per
17	1,000 between providers. So there is a
18	facility-level variation in quality.
19	And if you extrapolate that to
20	national data that's about 27,000 hospital
21	visits following colonoscopy procedure
22	nationally.

	Page 292
1	And that's really important
2	because most of these procedures, we know two-
3	thirds to three quarters are screening
4	colonoscopies. So by definition these are
5	procedures occurring in relatively healthy
6	people who don't have signs or symptoms of a
7	disease. And ensuring monitoring quality in
8	that group is there's a strong mandate for
9	measuring quality.
10	We also know when we look at the
11	top diagnosis that many of these patients come
12	back in with serious and very mild things, and
13	potentially preventable things. Things like
14	abdominal pain, nausea, bloating, bleeding,
15	perforation, syncope, aspiration because of
16	the anesthesia. So we think these are really
17	important things to measure and potentially
18	preventable.
19	The key thing with this is many of
20	these providers in the outpatient setting are
21	completely unaware of these events. They're
22	simply invisible.

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	Page 293
1	So for example, there's a study
2	from Boston that suggests that about 80
3	percent of these visits the providers are
4	actually unaware of. And you can understand
5	that. So if you're an ASC, ambulatory
6	surgical center, you're legally not allowed to
7	provide follow-up care. You can understand
8	why patients would present to a different
9	provider in the event of an adverse event. So
10	we think this measure is really important for
11	illuminating quality.
12	And if I can finish by very
13	quickly saying that this measure was developed
14	with input from working group that we had
15	input from Dr. Ron Bender and John Allen who
16	are heads of the American Gastroenterology
17	Association and the American College of
18	Gastroenterology.
19	We were extremely fortunate to
20	have them and they had a huge input into
21	shaping this measure.
22	We were also very lucky to have

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	Page 294
1	the highly qualified technical expert panel
2	which provide us great insights into this
3	measure. And a period of public comment which
4	we addressed many of the issues that came up.
5	And I must say overwhelmingly this
6	measure has been supported by that group. So
7	we think that provides a lot of rationale.
8	This measure uses Medicaid data so
9	it's eminently feasible. It's risk-adjusted.
10	It uses a hierarchical model. We can
11	statistically determine the outliers.
12	And just lastly, we submitted this
13	measure for approval to MAP during the
14	development process and we have received
15	conditional approval from them.
16	CO-CHAIR HALL: And just
17	immediately clarifying, the accountable entity
18	is the outpatient department or the ambulatory
19	surgery center?
20	DR. RANASINGHE: That's right.
21	CO-CHAIR HALL: All right. So,
22	thank you for that introduction to the

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	Page 295
1	measure. We're in the category of evidence.
2	We'll ask Ross and/or Cristie if they would
3	like to open the discussion.
4	DR. EDMUNDSON: Sure, I can start
5	here if that's okay with you, Cristie?
6	MS. TRAVIS: Please.
7	DR. EDMUNDSON: Okay. On
8	evidence, so you've already alluded to some of
9	the evidence. But there's not a lot. This is
10	a new measure.
11	I think you mentioned four
12	reasons why this is important. I'd add a
13	fifth one. When I need my colonoscopy I don't
14	want a complication. And that's important to
15	everybody because this is our general
16	population. If you live long enough you
17	should have one of these.
18	So the evidence that is literally
19	reviewed shows complication rates from 20 to
20	34 percent. And then you did have on HCUP
21	data unplanned hospital visits ranging from
22	8.2 to 20.1 per 1,000 colonoscopies.

	Page 296
1	DR. RANASINGHE: That's right.
2	DR. EDMUNDSON: And I think that -
3	- so there's some evidence out there although
4	it's light at this point in time for the
5	evidence.
6	Cristie?
7	MS. TRAVIS: No, just that you all
8	did document at least some of the
9	interventions that you thought would be
10	possible to actually improve upon the measure.
11	So from my perspective it meets the evidence
12	for health outcome.
13	CO-CHAIR HALL: Any other comments
14	or concerns? We'll consider this an
15	intermediate outcome for now. No other
16	comments. No cards. Let's vote on evidence.
17	MS. SHAHAB: Voting for 1(a)
18	evidence. One is yes, two no. Time starts
19	now. One more vote.
20	All votes are in for 1(a)
21	evidence. Fourteen yes, four no.
22	CO-CHAIR HALL: Performance gap

Page 297 1 opportunity. MS. TRAVIS: There does appear to 2 be variability in performance. 3 The standardized range was from 8.3 to 20.1. So 4 there seems to be quite an opportunity for 5 6 improvement. 7 CO-CHAIR HALL: Additional 8 comments? 9 DR. EDMUNDSON: I agree. No other 10 comment. 11 CO-CHAIR HALL: Let's move to 12 vote. 13 MS. SHAHAB: Voting for 1(b) performance gap. One high, two moderate, 14 three low, four insufficient. Time starts 15 16 now. 17 CO-CHAIR HALL: Let's try again. MS. SHAHAB: All votes are in for 18 1(b) performance gap. Seven high, eleven 19 moderate, zero low, zero insufficient. 20 21 CO-CHAIR HALL: Priority. MS. TRAVIS: This is a U.S. 22

Page 298 1 Preventive Services Task Force recommendation as was talked about since we're recommending 2 3 that people go get this as a screening test that we have an obligation to be sure they're 4 getting a high-quality. I think I was 5 impressed by the 14 million colonoscopies that 6 were done back in 2004 alone. So a very high-7 8 frequency procedure that needs to have the quality measured. 9 10 CO-CHAIR HALL: Additional 11 No additional comments. comments? MS. SHAHAB: Voting for 1(c) high 12 13 priority. One high, two moderate, three low, four insufficient. And your time begins now. 14 All votes are in for 1(c) high 15 16 priority. The results are 12 high, 6 17 moderate, zero low, zero insufficient. CO-CHAIR HALL: 18 Thank you. Moving into scientific acceptability, reliability. 19 20 DR. EDMUNDSON: The reliability 21 was on a 2010 split population arm. Large numbers of colonoscopies. 22

	Page 299
1	And I believe your the ICC on
2	that was 0.335 judged as fair. And I think
3	that's the information provided. Is that
4	correct?
5	DR. DRYE: Correct. We did, as
6	you heard on our other measures, we have
7	recalculated that with the Spearman-Brown
8	prophecy formula. It's an interesting name.
9	And it's at 0.43. That gives us a better
10	estimate if we had a full data set.
11	I would just note I think as
12	you're all aware the outcome rates for this
13	measure are lower than they are for, for
14	example, AMI readmission. And so we have to
15	get a bigger sample size to get reliable
16	measure score results.
17	And we're happy with what we're
18	seeing here but this is an inherent challenge
19	in a healthier population with a high volume
20	of procedures but a lower outcome rate.
21	CO-CHAIR HALL: Any additional
22	comments or concerns on reliability? Seeing

	Page 300
1	none.
2	MS. SHAHAB: Voting for 2(a)
3	reliability. One high, two moderate, three
4	low, four insufficient and the time starts
5	now. We need one more vote, please.
6	All votes are in for 2(a)
7	reliability. The results are 1 high, 17
8	moderate, zero low, zero insufficient.
9	CO-CHAIR HALL: Moving into
10	validity. Opening remarks.
11	DR. EDMUNDSON: Okay, on validity
12	you had your technical expert panel that drove
13	a lot of I think the validity questions as I
14	read through your information.
15	In addition, you had split sample
16	two years of data. And on that C statistics
17	of 0.67. And your conclusions were that this
18	is good model discrimination.
19	The other piece of information
20	that you provided that I'd like to have you
21	comment on is the Charleston model. Better
22	than Charlson model in that I believe this is

	Page 301
1	your risk stratification tool. And the
2	Elixhauser model of your C statistics actually
3	being better than those as a risk model.
4	Would you comment, please?
5	DR. RANASINGHE: So, this is
6	Ranasinghe here again.
7	So, we did two things. One was
8	that we compared the C statistics between
9	development and the validation sample, and
10	then again in using data. So we developed our
11	data using the 2010 sample and then we
12	validated it again in the 2011 sample.
13	The other validation step we did
14	was to construct the risk adjustment model
15	using we constructed a risk adjustment
16	model based on our conceptual and statistical
17	understanding of what we total predict
18	hospital visits in this population.
19	But we wanted to benchmark against
20	a risk model that's already being used. And
21	one option was to use the Charlson and the
22	Elixhauser models which are widely accepted.

	Page 302
1	Please keep in mind Charlson and
2	Elixhauser are unknown to gastroenterologists,
3	specifically not colonoscopy-specific. In
4	fact, the Elixhauser model was to predict
5	mortality from what I understand.
6	But this would give us a good idea
7	of where our model sits and our C statistics.
8	And the model characteristics actually ended
9	up being better than both those models.
10	DR. DRYE: This is Elizabeth Drye.
11	I would just add that, again, we're in a novel
12	data environment in a novel setting. We don't
13	have other, you know, there are not like 5 or
14	10 other readmission measures, similar
15	measures we can compare it to.
16	So even though we were hoping with
17	thoughtful variable selection with a lot of
18	clinical input that we would do better than
19	these indices we felt we should at least use
20	some other approach to make sure we were
21	getting what we thought we should be getting.
22	CO-CHAIR HALL: Karen?

	Page 303
1	DR. JOYNT: I just have a few
2	additional questions. One is I know there's
3	not a lot of other quality measures in this
4	group. But is there anything else you can
5	compare this to? Do we know if procedure
6	volume matters for colonoscopy? The rate of
7	detection of abnormalities I know has been
8	proposed as a quality. Is there anywhere else
9	to sort of externally validate whether or not
10	the rates that you're seeing are indicative of
11	bad procedure as opposed to sick patient?
12	DR. RANASINGHE: It's Isuru here
13	again. It's a great question and one that we
14	find really challenging because really there
15	is no outcome measures full stop for any of
16	the ambulatory measures. And we didn't really
17	know what to compare against.
18	And in fact we our conclusion
19	we reached was that there is no measure that
20	we could adequately compare against.
21	The only sort of measure that
22	comes close is the NQF measure that NQF-

Page 304 1 approved measure for ambulatory surgical centers which does an immediate transfer 2 3 following the procedure. But at the time we didn't have the actual reports of those 4 Individual hospital-level values were values. 5 not reported for that measure for us to 6 compare against. 7 DR. DRYE: I would just add also 8 9 that -- Elizabeth Drye again -- we relied 10 heavily on looking at the reasons for the ED 11 visits, observation stays and readmission and just thinking clinically, you know, are these 12 13 likely to be just sick patients, or are they, you know, did they look like they're related. 14 And also we know we're dealing 15 16 with a patient group and a procedure that 17 typically would not be done in an outpatient setting on patients who were acutely ill for 18 other reasons. 19 We did pull out of the measure 20 21 those patients who we felt might end up in the hospital for the reason for which they were 22

	Page 305
1	having the colonoscopy. So, IBD, inflammatory
2	bowel disease patients, patients with history
3	of diverticulitis are pulled out of the
4	measure. We had a lot of discussion around
5	that and comment on that issue.
6	We also pull out of the outcome
7	admissions for planned care like colorectal
8	resection. And that's one-third of all
9	hospital admissions that we see following
10	colonoscopy. So we tried to triangulate or
11	whatever, get at that concern in a number of
12	different ways.
13	We also looked at the baseline
14	admission rates and we looked at the falloff
15	in the first few days to try to pick the
16	outcome time frame. And we're confident that
17	we are zeroing in on hospital visits that are
18	really related to the procedure versus patient
19	factors that we're not adjusting for. But it
20	took a lot of different strategies.
21	CO-CHAIR HALL: Sherrie?
22	CO-CHAIR KAPLAN: I was just going

	Page 306
1	to sort of notice, and this is not for slowing
2	the conversation down or anything, but
3	different from the dialysis measure where the
4	attribution was back to the hospital, right,
5	for readmission to the hospital.
6	CO-CHAIR HALL: No, it was to the
7	dialysis.
8	CO-CHAIR KAPLAN: It was to the
9	dialysis center. So then it does follow the
10	same issue.
11	So the skill in that case of the
12	colonoscopist so confounded with center. At
13	that time there was some discussion about the
14	interest of the physician-level stuff and the
15	potential for estimating things, or
16	attributing things to the outpatient center
17	versus the providers.
18	And there was some conversation
19	around that. And I didn't want to I don't
20	want to cause a stir up a storm here, but
21	I did want to kind of for fairness of
22	comparison raise that issue.

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	Page 307
1	CO-CHAIR HALL: Actually in a
2	similar place because we're assuming that the
3	facility has some impact on quality and it's
4	not just the operator's skill and proficiency.
5	DR. DRYE: That's a great issue.
6	I think it's come up before.
7	The reason that we put the measure
8	at the facility level, there's a couple of
9	reasons.
10	One, there is a component of
11	facility care that we think contributes to the
12	outcome. There's the anesthesia care, there's
13	the post-op care, there's decision about when
14	a patient is ready to go home.
15	These facilities, a lot of them,
16	particularly ambulatory surgery centers are
17	physician-owned and they tend to specialize
18	and they may have a couple of physicians who
19	regularly work there.
20	But in general the physicians do
21	colonoscopies in multiple settings too. So if
22	you wanted to look at volume at the facility

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	Page 308
1	related to outcome you might not necessarily
2	be getting at the volume that you were trying
3	to capture which would be the per-physician,
4	at least for the physician component.
5	We're confident given the nature
6	of things that we see people go back to the
7	hospital for that there is a facility
8	component. And we expect that there's a
9	consistency at which physicians are doing the
10	care.
11	Also, as I mentioned before we
12	need to get a certain volume of patients and
13	outcomes to be able to get a reliable
14	estimate. So for all those reasons we housed
15	it at the facility level.
16	Like the other measures this
17	measure would be, if it's implemented by CMS
18	would be implemented in the hospital
19	outpatient department. I mean, sorry, the
20	hospital outpatient prospective OPPS system
21	as well as under the ambulatory surgery center
22	program.

	Page 309
1	So, they have not had this type of
2	measure as far as I know implemented there.
3	And we have encouraged CMS to follow the
4	approach that is taken for the hospital-based
5	measures in which facilities get patient-level
6	data, they can see which patients are in the
7	measure and where they ended up which are
8	things they won't otherwise see. And we think
9	that's going to be really critical.
10	And they won't be able to see in
11	there who the physician was, who the
12	anesthesiologist was, et cetera.
13	CO-CHAIR HALL: Paula.
14	MS. MINTON-FOLTZ: Does this
15	include immediate admission to either obs? Or
16	is there is a 24-hour lag?
17	DR. RANASINGHE: So this includes
18	sort of direct admissions from ASCs or from
19	HOPDs. Hospital outpatient departments.
20	And the rationale was that this is
21	colonoscopy procedures should be
22	straightforward procedures, range from 30

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	Page 310
1	minutes to an hour. There's really no reason
2	for the patient to stay for an extended length
3	of time unless some sort of adverse event
4	occurs.
5	MS. MINTON-FOLTZ: Well, there are
6	those patients who have no ride even. We've
7	seen those. But it's not a large percent.
8	CO-CHAIR HALL: Frank?
9	DR. BRIGGS: I was wondering if
10	you had data in regards to the how did you
11	come up with the seven days, seven days for
12	this type of procedure. And the side effects
13	that you're describing trying to capture seven
14	days actually seems a little bit long. And
15	you're thinking the first day, two days, maybe
16	three days. Out seven days you're probably
17	looking at continuation of symptoms and things
18	like that. So I was wondering if you had data
19	specifically to support your cutoff of seven.
20	DR. RANASINGHE: So that's a great
21	question. So there's a range of side effects
22	that could occur after a colonoscopy. And

	Page 311
1	they range from within a few days for things
2	like perforation but up to 30 days for things
3	like bleeding, GI bleeding. And so there's a
4	definite phenomenon of delayed GI bleeding.
5	We know from the literature that a
6	vast majority of those complications or
7	adverse events occur within the first seven
8	days.
9	And we can empirically test that
10	by looking at the number of hospital visit per
11	each day post procedure. And what we see is
12	a curve which sort of levels off to after
13	about seven days. And that's why we picked
14	the seven-day time window because we thought
15	that would give us the best sort of quality
16	signal for our measure.
17	So it doesn't and that was
18	supported by our technical expert panel. It
19	does mean that we miss some bleeding events
20	that are delayed. But we specifically
21	excluded them because we did not want to
22	capture hospital events that aren't related to

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	Page 312
1	the procedure.
2	CO-CHAIR HALL: Wes?
3	DR. FIELDS: Yes, a couple of
4	comments to Sherrie's point and then one
5	question.
6	The comments. Hopefully we'll see
7	many outpatient measures in the future. I
8	think this one actually has a pretty elegant
9	design.
10	But to Sherrie's analogy one of
11	the things I like about this is that they work
12	pretty hard at the development level to
13	identify what's essentially a well population.
14	So that's quite different than the standard
15	dialysis patient who's by definition
16	chronically ill.
17	The other is this is pretty much
18	of a 1 to 1 linkage between the physician
19	doing the colonoscopy than the procedure and
20	the outcome. And that's not necessarily true
21	of dialysis centers in terms of who is
22	actually providing the service and whether

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	Page 313
1	it's a passive or active physician activity.
2	But one question about the design
3	measure. I'm just sort of curious why you
4	chose not to include unscheduled follow-up
5	with a primary care physician or clinic during
6	the first seven days.
7	DR. RANASINGHE: So that's a great
8	question. So, it's very, firstly, we
9	considered all outcomes that are possible.
10	We felt the acute care user or
11	visit to a hospital is something that's
12	unexpected following a colonoscopy procedure.
13	And so that we thought would reflect a clear
14	quality signal.
15	Whereas visits to a primary care
16	provider could be planned, could be
17	appropriate care, could be scheduled care.
18	And that's very hard for us to identify from
19	claims data. And that is I guess the primary
20	reason.
21	CO-CHAIR HALL: You okay with
22	that, Wes?

	Page 314
1	DR. FIELDS: Oh, I'm very okay
2	with it. And for Helen and many other people
3	around the table I think the measure
4	developers make the point that there are many
5	situations where a component of primary care
6	or first contact care can happen someplace
7	besides a primary care clinician's practice.
8	And that's part of why I'm
9	fundamentally uncomfortable along with 30,000
10	of my close friends in emergency medicine with
11	the implied negative metric that goes along
12	with ED visits in many of the measures.
13	CO-CHAIR HALL: Point well taken.
14	Ross?
15	DR. EDMUNDSON: Yes, thanks for
16	bringing up the subject here.
17	I wrestled with this a lot because
18	I was one of the telephonic calls saying well,
19	my first impulse was this should be attributed
20	to a physician.
21	And as I wrestled with it I
22	thought, no, I have to address the methodology

	Page 315
1	that's in front of me. And I think there's
2	very good value to knowing what particular
3	facility had what kind of outcomes of
4	increased visits into the emergency room or
5	admissions.
6	I think there's value there. I
7	personally though, my bias is that if you did
8	the study and you attributed it, it's the same
9	information. And I agree with you it's a 1 to
10	1 relationship with a provider, a physician at
11	this point in time.
12	That physician is the one who will
13	meet face to face with the patient beforehand,
14	gives the instructions, gives the prep, does
15	the same prep and in fact will very often say,
16	well, where would you like to have this done.
17	We can do it in the hospital, you
18	can do it in my center or you can do it at
19	this based on their whim, their
20	preferences, the day of the week, the
21	convenience to the physician.
22	So, I think it would be a better

	Page 316
1	measure of quality tied to the physician. But
2	I do think it does have as it's stated and
3	presented to us some value as a facility as
4	well.
5	CO-CHAIR HALL: Thank you, Ross.
6	Karen?
7	DR. JOYNT: Just a quick question.
8	First, I commend you for again trying to reach
9	into the outpatient setting and look at
10	something that I think is probably under-
11	studied because of the difficulties like that.
12	Is there anything we should be
13	thinking about differently? Because this is
14	a composite of things that are very different.
15	Sort of a quick visit to the ED versus an
16	observation stay versus a hospital admission.
17	Or are these do you think, given
18	the healthy population that you've selected we
19	should consider them all to be within seven
20	days equivalent events?
21	DR. DRYE: So that's a great
22	question. We like to think of them as like

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	Page 317
1	they're above a certain threshold of acuity.
2	You feel bad enough that you have to go to the
3	ED.
4	And they may reflect different
5	things. They may reflect that you can't get
6	into your primary care doctor's office because
7	they're just not accessible to you the next
8	day or that evening or whatever.
9	And so we're just saying is this a
10	threshold effect, either because of the
11	serious problem or because of a problem that
12	could have been prevented, or could have been
13	cared for more efficiently in an outpatient
14	setting. That's where the patient is going.
15	Are they all equal? They're not
16	all equal. And I think, you know, again we
17	feel really strongly the data has to be
18	reported back at the patient level with the
19	reason and the location of the hospital visit
20	so that providers can look at that.
21	I think, my own view is if three,
22	you know, gastroenterologists or general

	Page 318
1	surgeons are at an ambulatory surgery center
2	doing this care and one is having outcomes
3	that are really driving those scores, that's
4	a lot of peer review and direct pressure back
5	on one provider.
6	So, but it gets back to your point
7	about how things get put into use. I think
8	with that information available to facilities
9	this works well as a composite and it gives us
10	the volume that we need.
11	CO-CHAIR HALL: Thank you. Ross,
12	another question? You all right? I have a
13	possibly small question but I'm curious as to
14	why you risk-adjusted for polypectomy.
15	DR. RANASINGHE: So that's a great
16	question. We know from the literature that
17	polypectomy is associated with bleeding, GI
18	bleeding, and that's the strongest risk factor
19	for bleeding.
20	At the same time there's a lot of
21	debate about, you know, removing the polyp
22	could be discretionary, that some providers

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	Page 319
1	may remove unnecessary polyps and you might be
2	you're adjusting away that quality signal.
3	But we felt removing
4	identifying polyps and removing them is a
5	quality indicator itself of colonoscopy. We
6	did not want to you know, if somebody
7	appropriately removes a polyp and that
8	resulted in a high rate of hospital visits we
9	didn't want to disadvantage these providers
10	who are taking appropriate action to treat a
11	condition.
12	CO-CHAIR HALL: I understand that,
13	but you could also argue that there's perhaps
14	proficiency involved in taking polyps out
15	without causing bleeding.
16	DR. RANASINGHE: I just want to
17	make clear that we do not adjust for the
18	technique used in removing the polyp, the
19	number of polyps. There's a number of
20	specific techniques for removing polyps. So
21	that technical component we do not adjust for.
22	And that is up to the discretion of the

	Page 320
1	provider.
2	CO-CHAIR HALL: I understand. I
3	guess it looks to me like a part of the
4	therapy that could cause the event, the
5	complication, and you may be adjusting for it.
6	Any other? Paul?
7	DR. HEIDENREICH: Well, just it
8	sounds like that should be the next measure,
9	a companion measure of polypectomy rates for
10	those undergoing polypectomy.
11	CO-CHAIR HALL: Or adenoma
12	detection rates perhaps. Other comments in
13	the category of validity? Seeing none.
14	MS. SHAHAB: Voting for 2(b)
15	validity, one high, two moderate, three low,
16	four insufficient and time starts now.
17	All votes are in for 2(b)
18	validity. The results are zero high, 18
19	moderate, zero low, zero insufficient.
20	CO-CHAIR HALL: Feasibility.
21	Opening remarks? Anyone? Sorry, go ahead,
22	Cristie.

	Page 321
1	MS. TRAVIS: Administrative claims
2	and therefore feasible to collect.
3	CO-CHAIR HALL: Other comments?
4	Seeing none.
5	MS. SHAHAB: Voting for
6	feasibility. One high, two moderate, three
7	low, four insufficient and the time starts
8	now.
9	All votes are in for feasibility
10	and the results are 14 high, 4 moderate, zero
11	low, zero insufficient.
12	CO-CHAIR HALL: Usability.
13	Opening remarks? Ross or Cristie, any
14	specific remarks?
15	MS. TRAVIS: Nothing specific.
16	It's just not in use yet. But they did talk
17	about the fact that if it was publicly
18	reported or later used in payment
19	methodologies and talked a little bit about
20	using the same methodology as for some of the
21	other CMS measures.
22	CO-CHAIR HALL: Kathy?

	Page 322
1	DR. AUGER: I just wanted to
2	comment on a potential unintended risk. As
3	the developer mentioned a few minutes ago if
4	you're in a group of three people at an
5	ambulatory care practice and one of those
6	providers is really driving the outcome and
7	you end up in an outlier as perhaps a bad
8	performing site.
9	As a member of the public knowing
10	that might actually give you anxiety about
11	your physician's ability which may be
12	completely misattributed. You might actually
13	be wrongfully attributing risk to a good
14	doctor as opposed to one that has higher
15	adverse outcomes. So, there's the potential
16	for misattribution of risk. But I suppose you
17	could also argue that you could then put
18	pressure on that lower performing physician as
19	well. So I could go either way.
20	CO-CHAIR HALL: Karen.
21	DR. JOYNT: I may just be reading
22	this wrong so feel free to correct me if I'm

Page 323 1 mistaken. But in the long report it looks 2 like from the HCUP data that one facility was 3 found to be better than expected and four were 4 found to be worse than expected. 5 And I feel like from what you told 6 us at the beginning about the variability in 7 outcomes that that either strikes me as the 8 9 sample size is too small or I misunderstood 10 the variability. Or this isn't the same you 11 were talking about. So just a clarification on sort of how this plays out in the real 12 13 world data would be helpful. Okay, that's a 14 DR. RANASINGHE: great question. So, we developed our measure 15 16 using a 20 percent Medicare sample. And 17 because that's a sample we needed to actually test the measure score. We need a 100 percent 18 19 sample. 20 And for that we used HCUP data 21 from four states. And that outlier analysis is data from four states only. And that's 22

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	Page 324
1	about 992 facilities. So we expect many more
2	outliers. If you actually included the
3	national population I think there's about
4	8,000 facilities plus nationwide.
5	The other point I need to make is
6	that that is using the 95th percentile and
7	that's a policy decision. So if you wanted to
8	capture more outliers you can change the
9	cutoff interval.
10	CO-CHAIR HALL: So with the 5/95
11	interval you're working at about a percent, 1
12	percent roughly more or less? Okay.
13	Other comments? Usability, other
14	comments or concerns? Not seeing any.
15	MS. SHAHAB: Voting for usability
16	and use. One high, two moderate, three low,
17	four insufficient information and the time
18	starts now.
19	All votes are in for usability and
20	use. And the results are 1 high, 16 moderate,
21	1 low and zero insufficient information.
22	CO-CHAIR HALL: Any summary
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	Page 325
1	comments for overall voting? I do not see
2	any. Ross?
3	DR. EDMUNDSON: I think this is a
4	very a high-frequency outpatient procedure.
5	And this is a start. This is shining the
6	light on it.
7	And I think that there's a lot
8	that we're going to probably have to learn out
9	of this one. And I think like as I say, my
10	prejudice is that I think it will be better to
11	look at this from an individual provider. But
12	I'll leave that for time and further
13	information to decide.
14	CO-CHAIR HALL: Thank you. Any
15	other comments. No? Let's move to vote.
16	MS. SHAHAB: Voting for overall
17	suitability for endorsement, one yes, two no.
18	The time starts now.
19	CO-CHAIR HALL: We're waiting for
20	one more if you could just click it one more
21	time.
22	MS. SHAHAB: All votes are in for

Page 326 1 overall suitability for endorsement for measure 2539 Facility 7-day Risk-standardized 2 3 Hospital Visit Rate after Outpatient Colonoscopy. And the results are 17 yes, 1 4 5 no. CO-CHAIR HALL: We thank the 6 7 developers for their effort and input. And we will move into the review 8 9 of 1789. I'm turning over to who? Who am I 10 turning over to? 11 MR. AMIN: Okay, so --CO-CHAIR KAPLAN: Before you 12 13 start, is part of your start going to be what we're supposed to do here? 14 15 MR. AMIN: Yes. 16 CO-CHAIR KAPLAN: Thank you. 17 (Laughter) CO-CHAIR HALL: And we welcome the 18 Yale team back to the table. They'll 19 introduce themselves in a moment. 20 21 Okay. So, as part of MR. AMIN: the 2011 evaluation of measure 1789 the 22

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	Page 327
1	Hospital-wide All-cause Unplanned Readmissions
2	Discussion there was a number of elements that
3	the steering committee requested from the
4	developer post dry run, especially trying to
5	understand the results of the dry run,
6	specifically an analysis of the distribution
7	of performance between hospitals with varying
8	proportions of low SES patients and the
9	proportion of the measure result variation
10	that is attributable to providers compared to
11	patients.
12	This information was not available
13	during the initial endorsement of this measure
14	and so given that we this is the first time
15	this committee has met since the evaluation of
16	1789 we've asked the measure developers to
17	provide this information to the committee.
18	And I will turn it back over to
19	and provide a quick update on the progress
20	related to harmonization to measure 1768 which
21	was provided to members of the committee that
22	are returning.

	Page 328
1	So there are a number of members
2	of the committee that are aware of this issue.
3	And I would sort of invite you to participate
4	in this conversation.
5	I'll just note who that is.
6	Bruce, Cristie, Jo Ann Brooks, Paula, Larry
7	Glance, Leslie Kelly Hall and Sherrie Kaplan
8	of course.
9	So, that is the topic of the next
10	half hour.
11	CO-CHAIR HALL: Thank you. And
12	just to clarify, we will not actually be asked
13	to vote on anything.
14	MR. AMIN: No.
15	CO-CHAIR HALL: We're being asked
16	to review the information.
17	MR. AMIN: Yes, these are purely
18	updates from the developer and conversations
19	that the committee may want to have with the
20	developer related to those topics. But this
21	measure is not up for review and there will
22	not be an endorsement decision from this

Page 329 1 conversation. CO-CHAIR KAPLAN: And the goal of 2 3 this conversation is to generate? Is to provide an update 4 MR. AMIN: to the committee. This was --5 CO-CHAIR KAPLAN: Not on their 6 7 side, on our side. What are we to do? We're 8 to provide what? 9 MR. AMIN: This was a request by 10 this committee at the end of the last review. 11 So this is an update on these particular issues. There's no action required by the 12 13 committee at this point. DR. BURSTIN: And just to add to 14 So, when the decision was made to 15 that. endorse this measure there was -- some of you 16 17 may remember this was not without controversy. Susannah is still smiling, that's good. 18 But part of the agreement 19 20 particularly with the NQF board as well as 21 with the steering committee was that they wanted us to take a look back to see what the 22

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	Page 330
1	experience has been, to see if there is any
2	evidence of unintended consequences and sort
3	of monitor the situation. This is essentially
4	that monitoring update.
5	CO-CHAIR KAPLAN: So, essentially
6	not to keep beating this horse, but
7	essentially our role is to thank you for
8	sharing and we are any comments we have for
9	the developer will be communicated to
10	MR. AMIN: They will be
11	communicated in the report around the updates
12	on these topics.
13	DR. BURSTIN: And certainly since
14	the developers and CMS are here we want them
15	to be part of this discussion. If there are
16	issues that are raised that need further
17	discussion we will encourage those
18	discussions.
19	CO-CHAIR KAPLAN: Could the
20	developers please re-introduce yourselves for
21	the record and then present your findings.
22	DR. HOROWITZ: I'm Leora Horowitz

	Page 331
1	from Yale.
2	DR. BERNHEIM: And I'm Susannah
3	Bernheim.
4	DR. HOROWITZ: So I led the
5	development of this measure and presented to
6	this committee in 2011. And thank you. I
7	know it's been a long two days so thanks for
8	bearing with us.
9	So, as you know this measure was
10	endorsed in the spring of 2012 and at that
11	time you had asked us to come back earlier
12	than the three years to talk about the dry run
13	and the harmonization. And so I'm going to
14	just quickly summarize those. And I believe
15	you've received those materials as well.
16	So, we had the dry run in the fall
17	of 2012. Dry run means that CMS sends
18	hospitals a confidential report of what the
19	measure results look like but without publicly
20	reporting them.
21	And so hospitals received the
22	overall score. They also received the scores

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	Page 332
1	for each of the five specialty cohorts that
2	make up this measure. And they received a
3	list of every one of their patients that was
4	in the measure along with whether that patient
5	had been readmitted and if so to which
6	hospital and what date and what the procedures
7	and the diagnoses were.
8	And there was a lot of interest in
9	that dry run. CMS sent results to 4,652
10	hospitals. Seventy-two percent of them
11	downloaded their data. We received 163
12	questions about the measure after that dry run
13	and we had 2,400 people approximately register
14	for two phone calls that we had to explain the
15	measure and answer questions.
16	Most of the questions were about
17	the methodology in various ways.
18	So based on the feedback and the
19	questions that we got in that dry run we made
20	several changes to the measures. And I want
21	to make sure that you're aware of those now.
22	We updated our planned readmission

	Page 333
1	algorithm. At that time we added nine
2	procedure categories to the list of things
3	that could qualify you as having a planned
4	readmission. We also removed a diagnosis code
5	from the acute diagnosis list which would have
6	otherwise disqualified a readmission from
7	being called planned.
8	And those changes were already
9	brought before NQF in the context of the re-
10	endorsement of our other various other
11	condition-specific measures.
12	For this measure, for the
13	hospital-wide measure those changes increased
14	the proportion of readmissions that we called
15	planned. So before we called 5.1 percent of
16	readmissions planned and with the changes we
17	called 8.3 percent of them planned. So that
18	decreased the national unplanned hospital
19	readmission rate from 16.8 to 16.2 percent.
20	I should add that we subsequently
21	conducted a chart validation of the planned
22	readmission algorithm and have made several

	Page 334
1	more small changes that have also slightly
2	changed the readmission rate. But those
3	changes are smaller.
4	Based on the feedback that we got
5	in the dry run we also altered the assignment
6	of some patients from the surgical cohort to
7	other cohorts, so about 200,000 patients
8	overall moved. And we changed the way the
9	unplanned readmission following a planned
10	readmission was counted in the measure.
11	Again, that change has been brought before NQF
12	for other measures.
13	So, with regard to harmonization
14	we did include in our materials an updated
15	memo from both us and NCQA. NCQA is the owner
16	of the Plan All-cause Readmission measure
17	which is another all-cause readmission measure
18	but targeted at the health plan level, not at
19	the hospital level.
20	And at the time when we originally
21	had the endorsement we were asked to talk
22	about eight different areas in which we were

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	Page 335
1	not harmonized to see if we could harmonize on
2	them.
3	So, one it turned out we were
4	already harmonized on so that was easy.
5	Two more NCQA is planning
6	hopefully to harmonize on. But they have put
7	those decisions out to public comment and then
8	it needs to be voted on by their board of
9	directors. So their plan is to harmonize on
10	two other topics which is the planned
11	readmission algorithm and counting
12	readmissions as new index admissions. And so
13	we will hear about that when that vote
14	happens.
15	For two other areas in which we're
16	not harmonized NCQA did several analyses.
17	Those are the exact form of the risk
18	adjustment variables that we used and using
19	hierarchical versus non-hierarchical modeling.
20	When they analyzed those two areas
21	they found it made almost no difference at all
22	to the measure. And so for the purposes of

	Page 336
1	just sort of practicality and simplicity we
2	are both going to continue with the approaches
3	that we take knowing it doesn't really matter
4	very much.
5	And then there are the remaining
6	issues that we have so far agreed to disagree
7	on still.
8	One is patients receiving medical
9	treatment for cancer. We exclude those
10	patients from the hospital measure. They have
11	extremely high post-discharge mortality rates
12	and we're worried about the competing risk.
13	On a plan level that's less
14	relevant to the plan population. NCQA prefers
15	to keep those patients.
16	The other area that we are going
17	to continue to disagree on is psychiatric
18	patients. We both include patients who are
19	admitted with substance abuse or other sort of
20	medical psychiatric problems.
21	But patients who are only admitted
22	for a psychiatric disease, like for an acute

	Page 337
1	schizophrenia or acute bipolar disorder are
2	not in our measure because we our measure
3	doesn't have psychiatric hospitals.
4	And so psychiatric hospitals and
5	psychiatric units are not in our measure.
6	They are in the NCQA, sort of the plan
7	measure.
8	And so there's a very small
9	fraction, about 3 percent of all psychiatric
10	admissions come into medical hospitals perhaps
11	because we just can't exclude them well. And
12	so we exclude that tiny fraction because
13	they're just a very small fraction of all
14	psychiatric patients. So we're going to
15	continue to disagree on that point.
16	And lastly with regard to
17	socioeconomic questions it's obviously
18	extremely complicated and difficult.
19	So after the original committee
20	meeting we put together a variety of analyses
21	trying to understand what the relationship of
22	hospital and patient kind of level variability

	Page 338
1	was.
2	And we sent those analyses to the
3	board after the committee had met. But the
4	committee never got a chance to see them. So
5	we included those analyses in the materials we
6	sent to you.
7	And the 30,000 foot view very
8	quickly is just that hospitals that have
9	higher rates of patients with lower
10	socioeconomic status which in itself is a very
11	hard thing to kind of get your brain around,
12	but we did that in four or five different ways
13	so we could try to capture that.
14	Those hospitals that have a lot of
15	those patients do have slightly higher risk-
16	standardized readmission rates. Although the
17	overlap is really profound so there's just a
18	huge amount of overlap in the rates.
19	And then but that didn't
20	particularly help us much because that could
21	have been because there's an intrinsic risk to
22	having low socioeconomic status. Or it could

1							
	Page 339						
1	have been because patients with low						
2	socioeconomic status may cluster in hospitals						
3	with lower quality. And so it's so hard to						
4	disentangle what that means.						
5	So, the last thing that we did is						
6	we just took all of those patients out as best						
7	we could. So, in one analysis we removed all						
8	Medicaid patients from the data. So we tried						
9	as best we could to take out those						
10	socioeconomic status patients altogether. And						
11	then we redid the analyses and still we find						
12	a slightly higher risk-standardized						
13	readmission rate in the hospitals that have a						
14	lot of those patients even though we're not						
15	putting them in the measure.						
16	And so again, it's hard to exactly						
17	know what that means but it's at least						
18	suggested that this is not purely a patient-						
19	level problem, that there's some component						
20	relating to the hospital.						
21	CO-CHAIR HALL: Thank you. Leora,						
22	would you mind just reminding us how you						

Page 340 1 you touched on it, but you didn't state it in detail, how readmission is considered or not 2 considered an index as well in your algorithm? 3 DR. HOROWITZ: So, in this measure 4 every readmission is newly considered an index 5 admission. And so a readmission counts as a 6 readmission, and then it also counts as an 7 index and we look to see forward if there's a 8 9 readmission after it. 10 CO-CHAIR HALL: Thank you. Any 11 questions from the group? Wes. DR. FIELDS: Just one small one. 12 13 I wasn't in the room for this prior discussion. So I'm not clear. Is this an 14 all-plan analysis or analysis of CMS 15 16 populations? 17 DR. HOROWITZ: This is endorsed for 18 and over but it's in use currently only 18 for Medicare patients. 19 20 DR. FIELDS: So when you said you 21 removed the Medicaid patients, you're talking about the dual eligible population? 22

	Page 341					
1	DR. HOROWITZ: Correct.					
2	CO-CHAIR HALL: But the removal					
3	was just a sensitivity test. It wasn't					
4	actually how it's yes, okay.					
5	CO-CHAIR KAPLAN: So, I was out of					
6	the room as well. Just but recalling these					
7	data, no matter that it looked like robust					
8	across disproportionate share hospitals,					
9	whether or not your status was a public					
10	hospital, proportion of Medicaid patients, and					
11	I forget what the fourth one was.					
12	But robust across about four or					
13	five different considerations of things that					
14	actually could be a proxy. We can argue about					
15	what they're actually a proxy for. But at					
16	least for robust across those treatments of					
17	potential differences in socioeconomic status.					
18	Non-random clustering by patients within					
19	hospital. Your findings are reasonably robust					
20	across pretty much everything you tried.					
21	DR. HOROWITZ: Yes. As Sherrie					
22	said, we defined SES in every way we could					

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	Page 342						
1	knowing we only had administrative data. And						
2	so the things that often people really care						
3	about, whether they are literate or are						
4	homeless or things like that, we just don't						
5	have that data.						
6	But what we did was we looked at						
7	proportion of patients that are dual eligible						
8	that have Medicaid at the hospital.						
9	We looked at whether the hospital						
10	was a safety net hospital which we defined as						
11	being more than a standard deviation above the						
12	state average for its dual eligible patients.						
13	We looked at whether the hospital						
14	was considered a disproportionate share						
15	hospital by the government. And we looked at						
16	whether the hospital was a public hospital.						
17	So, those are all ways we tried to get at						
18	whether a hospital was going to see a						
19	disproportionate number of low-SES patients.						
20	And of all of those the tightest,						
21	the most conservative definition we used was						
22	proportion of Medicaid. So we looked at						

Page 343 1 hospitals that had 30 percent or more of their patients having Medicaid. That was our 2 3 smallest, most extreme sample. We only had 300-something, 331 hospitals like that. 4 So that's where you're going to see the biggest 5 If we're going to see anything 6 differences. it should be in those hospitals. 7 8 CO-CHAIR HALL: Any other delving into any other geographic qualifiers? 9 10 DR. HOROWITZ: We did not look at 11 any other geographic differences? MS. MINTON-FOLTZ: Did you account 12 13 for no-pay or undocumented? 14 DR. HOROWITZ: So again, because we did this in Medicare data all of our 15 16 patients by definition have Medicare. And so 17 all we really had was the dual eligible Medicaid and Medicare patients. 18 CO-CHAIR HALL: Any other comments 19 20 or questions? I'll look back to our NQF 21 colleagues. Do we just thank the developers or is there any other? 22

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	Page 344					
1	MR. AMIN: Yes, that's it.					
2	CO-CHAIR HALL: Thank you for this					
3	update.					
4	DR. HOROWITZ: Thank you for					
5	having us.					
6	MR. AMIN: Okay, so we have two					
7	other items. We have public and member					
8	comment at 2:30.					
9	I just wanted to point out a few					
10	next steps. I'm just going to turn it over to					
11	Adeela to talk through some of the next steps.					
12	One of the immediate next steps					
13	that I want the committee to be aware of is					
14	that we have a call scheduled on May 16 from					
15	2 to 4. And we will use that call, the					
16	majority of that call to discuss the					
17	harmonization of competing measures					
18	discussion.					
19	So we have three sets of competing					
20	measures discussions that we're going to have.					
21	Just so that there's a few folks in the room					
22	that we're going to ask you to play a little					

	Page 345					
1	bit of a role during that conversation.					
2	Those that have been lead					
3	discussants for the measures, Wes Fields and					
4	Pam Roberts, for the two there's the 2505,					
5	the ED use within 30 days of home health.					
6	There are two measures that are related to					
7	this measure. And we'll send you a side-by-					
8	side table to give you a description of what					
9	they look like. But just so you're aware. So					
10	again, Wes and Pam, we're going to ask you to					
11	lead this discussion during the call on the					
12	16th.					
13	So 2505 relates to 0173 which is					
14	the acute hospitalizations for home health					
15	patients. And 0171 which is ED use post home					
16	health without the 30-day qualifier. And so					
17	we're going to have to have a conversation					
18	related to how these are related and whether					
19	we should be selecting one of the measures for					
20	endorsement.					
21	The other set of measures, Helen,					
22	we're going to ask you to take the lead on					

	Page 346					
1	since you were the lead discussant on is 2375					
2	the SNF all-cause readmission measures and the					
3	2510 also the SNF all-cause readmission.					
4	And then the third set, you know,					
5	there are a number of lead discussants.					
6	Bruce, Paul, Ross and John, you guys were all					
7	part of the discussions around the CABG					
8	readmission measures. So we'll send a side-					
9	by-side table related to those two measures.					
10	But we'll have to have a conversation related					
11	to potentially selecting a best in class.					
12	We'll also send along prior to					
13	that call a description of the decision logic					
14	of how we will go through the discussion					
15	around either selecting one as a best in class					
16	measure or potential harmonization. But the					
17	nature of that call will be to discuss					
18	harmonization or selecting a best in class					
19	measure.					
20	Again, we'll follow up with much					
21	more detail in an email with some descriptions					
22	of those measures and who's responsible. But					

	Page 347						
1	before we leave today I wanted to make sure						
2	that we at least had a little bit of, you						
3	know, these recommendations are contingent on						
4	the fact that we have a discussion around						
5	selecting or at least addressing the question						
6	of harmonization or best in class.						
7	I think there are a few questions						
8	on that topic so I welcome them.						
9	MS. SHIPPY: Weill we be asked to						
10	vote on the phone call?						
11	MR. AMIN: We will I don't know						
12	the answer to that yet. But likely we won't						
13	be we'll have to make a decision. So						
14	likely it will be through a follow-up						
15	SurveyMonkey and not voting on the call						
16	itself.						
17	Again, we'll follow up with a lot						
18	more detail of exactly what will be kind of						
19	expected during that conversation. But it						
20	will be more of a lead discussant on the						
21	measures that you've already reviewed for the						
22	committee. So there shouldn't be anything new						

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	Page 348						
1	there.						
2	So, Adeela, I'll turn it over to						
3	you in terms of summary and follow-up in terms						
4	of next steps.						
5	MS. KHAN: Sure. You can look for						
6	that email around Friday. You'll get it in						
7	your inboxes by then.						
8	Just a quick summary of the						
9	measures that we've gone through today. 2515,						
10	the Hospital 30-day All-cause Unplanned Risk-						
11	standardized Readmission Rate Following						
12	Coronary Artery Bypass Graft Surgery passed.						
13	2514 Risk-adjusted Coronary Artery						
14	Bypass Graft Readmission Rate also passed.						
15	2393 Pediatric All-Cause						
16	Readmission Measure passed.						
17	2414 Pediatric Lower Respiratory						
18	Infection Readmission Measure passed.						
19	2513 Hospital 30-day All-cause						
20	Risk-standardized Readmission Rate Following						
21	Vascular Procedures passed.						
22	0695 Hospital 30-day Risk-						

	Page 349						
1	standardized Readmission Rates Following						
2	Percutaneous Coronary Intervention passed.						
3	0505 Hospital 30-day All-cause						
4	Risk-standardized Readmission Rate Following						
5	Acute Myocardial Infarction Hospitalization						
6	passed.						
7	And 2539 Facility 7-day Risk-						
8	standardized Hospital Visit Rate after						
9	Outpatient Colonoscopy also passed.						
10	In terms of next steps after this						
11	meeting we're going to have our post-meeting						
12	call that Taroon mentioned. It's scheduled						
13	for May 16 2 to 4. You should have that on						
14	your calendars already. If you don't let me						
15	know right away.						
16	After the in-person meeting we're						
17	going to start writing the report. And we						
18	expect the report to go out to public comment						
19	in June, early June, June 6 through July 7.						
20	That will be about a 30-day public comment.						
21	We have to accommodate for the July 4 holiday.						
22	But it will be 30 days.						

Page 350 1 And we encourage all of you to pass the report around and get as many 2 3 comments as we can. We'll have a steering committee 4 call to review and respond to the comments. 5 That's July 30. Again you should have that on 6 your calendars. 7 8 In August we expect the measures 9 to go through the NQF member vote and to CSAC 10 followed by endorsement by the board in 11 September. And we will have a 30-day appeals 12 13 period we'll start in October. And we'll have exact dates for you once the time is closer. 14 That's all I have for today. I'11 15 turn it back to Taroon. 16 17 MR. AMIN: So, I would just say from the NQF team a profound thank you very 18 much to the committee for all of the hours 19 20 that you spent reviewing all of these measures, all of the workgroup calls that you 21 spent and obviously this very entertaining but 22

1						
	Page 351					
1	exhausting two days that I'm sure we've had					
2	here.					
3	And a particular thank you to the					
4	co-chairs who have led us through this on time					
5	across the two days. You've saved us a lot of					
6	work in terms of scheduling follow-up					
7	conference calls to review measures. So thank					
8	you to Bruce and Sherrie for all of your work					
9	here.					
10	And we're just I'll ask for					
11	some reflections from the chairs. But I also					
12	want to be cognizant that we have the 2:30					
13	public comment period. So thank you.					
14	CO-CHAIR HALL: Well, I will just					
15	briefly say that Sherrie and I are thrilled to					
16	preside over such a wonderful group of experts					
17	in these areas. So the privilege has been all					
18	ours, all mine. I always, always learn from					
19	this process so I'm always thrilled to take					
20	part. And being with such a great group is					
21	what makes it worthwhile.					
22	CO-CHAIR KAPLAN: Ditto.					

	Page 352						
1	MR. AMIN: So, Operator, I want to						
2	see if there are public comments on the phone.						
3	And we'll also take any public comments in the						
4	room.						
5	OPERATOR: Okay. If you'd like to						
6	make a public comment please press * then the						
7	number 1. There are no comments at this time.						
8	MR. AMIN: Are there any comments						
9	in the room? No. Okay.						
10	Again, thank you all very much.						
11	And we again, we appreciate all of your work						
12	on this. Look forward to the follow-up call.						
13	(Whereupon, the foregoing matter						
14	went off the record at 2:25 p.m.)						
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A	acceptability 22:22	<b>active</b> 64:15 125:11	226:7 227:6 246:1	301:14,15 335:18
<b>\$136</b> 137:18	24:6 107:7 108:19	313:1	287:11 297:7	administration
<b>\$360</b> 226:4	138:16 198:2	actively 40:2,6	298:10,11 299:21	177:15
<b>A-G-E-N-D-A</b> 4:1	251:8 272:22	125:9 131:4	303:2	administrative
	298:19	activity 313:1	address 39:15	19:8,20 20:2
<b>a.m</b> 1:9 6:2 129:9	acceptable 242:21	actual 20:7 24:11	109:10,18 113:10	71:19 72:7,7
129:10 220:18	256:8	27:18 145:2	113:20 157:19	78:19 79:5 110:13
abdominal 102:15	accepted 301:22	162:22 170:10	214:4 215:21	182:14 240:18,18
292:14	access 46:8 52:19	214:12 234:3	221:21 283:1	263:5,10 264:2
<b>ability</b> 8:20,22 20:8	145:16 169:13	252:17 274:22	286:2 314:22	275:4 321:1 342:1
43:9 48:10 64:4,5	181:16 182:21	304:4	addressed 34:13	administrator
72:3 74:13,13	accessible 317:7	acuity 218:8 317:1	55:11 186:14	119:9
84:11 169:5	accommodate	acute 4:19 38:3	210:4 218:12	admirable 257:20
203:16 227:19	112:7 349:21	151:18 266:15	294:4	admission 37:13
322:11 abla 6:16 0:15 10:0	account 106:16	268:20 288:8	addressing 35:13	120:16 141:8,18
<b>able</b> 6:16 9:15 10:9	158:18 169:7	313:10 333:5	347:5	177:18 178:9
13:12,22 15:15	175:10 203:21	336:22 337:1	Adeela 2:13 13:8	181:20 200:18
36:16 58:17,19	229:12 230:2,9	345:14 349:5	15:11 211:21	218:8,10 305:14
69:17 87:15	264:6 343:12	acutely 304:18	344:11 348:2	309:15 316:16
111:12 142:5	accountability	ad 269:4	adenoma 320:11	340:6
144:14 147:4 152:14 156:8	112:2,8 123:10	Adam's 69:10	adequacy 250:13	admissions 1:3
157:16,18 160:7	213:14 215:2	adaptive 9:15	adequate 26:14	146:18 181:22
164:10,12 165:10	216:3 251:2	add 16:17 52:18	198:10	217:16 218:16
166:8 172:4,13	accountable 187:18	57:7 61:16 62:11	adequately 203:21	269:20,21 291:2,3
179:8,12 183:21	294:17	85:16 89:13 90:10	303:20	305:7,9 309:18
186:2 189:9 193:4	accounted 51:21	91:13 97:19	adhere 11:6 12:8	315:5 335:12
194:2 205:12	201:15	119:22 134:15	adherence 174:8	337:10
216:19 251:9	accounting 193:21	149:8 183:13	<b>adjust</b> 117:13,14	admitted 118:5
258:16 259:15	229:3 264:3	185:15 194:11	119:17 168:14	120:11 152:15
308:13 309:10	accounts 197:6	215:12 250:14	177:9 319:17,21	336:19,21
abnormalities	accuracy 32:9	254:20 255:21	adjusted 107:13	adolescents 130:9
303:7	accurate 263:8	271:1 295:12	120:4 224:6 244:3	adult 131:13 132:1
<b>absence</b> 131:6	achieve 15:15	302:11 304:8	adjuster 180:14	132:5 133:12
absolute 105:8	58:17 69:4	329:14 333:20	adjusting 180:14	134:9 137:20
191:22	achieved 28:5	added 333:1	305:19 319:2	143:1 146:13
absolutely 33:22	76:16,17	addition 16:6 77:3	320:5	147:13 148:4
55:12,13 79:1	achieving 77:4	96:20 102:10	adjustment 20:7	150:20 160:16
146:20 160:21	acknowledge	131:14 191:18	50:5,9,19 51:2	171:21 172:19
166:5 285:9,11	143:21 154:22	222:19 230:2	58:17 59:20 75:19	177:16 192:5
abstraction 226:22	174:11	244:5 251:6	75:21 76:4 79:17	195:7 199:10,15
abundance 252:13	acknowledged	253:16 300:15	92:10,12 102:6	210:12 213:9
<b>abuse</b> 336:19	147:9 201:9	additional 52:1	116:20 132:20	adults 130:21 158:3
ACC 221:11,16,19	ACNS-BC 1:16	56:14 62:22 66:2	140:11 168:16	168:19,20 174:2
230:17 254:2	acquired 191:13	66:17 70:7 93:22	192:11 210:9	175:3
258:12	act 158:8	95:12 145:8	213:6 217:4 218:6	advance 268:18
ACC's 223:9	action 319:10	176:21 188:20	219:20 224:22	advantage 10:14
	329:12	215:21 225:9	225:4 263:22	19:21 40:12
	•	•	•	•

### 277:5 anxiety 322:10 advantageous 70:9 75:13 124:22 186:8 249:22 84:19 184:14 193:16 **all-plan** 340:15 268:11 282:13 anybody 11:10 advantages 147:20 223:19 320:21 Allen 293:15 338:18 104:15 106:12 Adventist 1:17 **AHRO** 130:12 Alliance 2:4 analogous 196:3 169:17 170:5 analogy 168:18 adverse 131:2 132:17 194:21 **allow** 60:21 160:2 181:10 202:4 197:4 312:10 291:5 293:9 310:3 259:12 266:9 278:21 311:7 322:15 **aimed** 175:21 **allowed** 38:18 analyses 232:14 **anymore** 32:11 **advice** 75:17 268:2 **aiming** 162:4 45:20 58:11 335:16 337:20 aortic 30:16 **advocate** 45:15.16 air 285:17 267:20 293:6 338:2,5 339:11 apart 81:14.20 apologies 58:6 46:18 100:20 Alan 3:7 129:20 allows 74:1 259:6 analysis 25:17 142:18 148:14 alluded 154:19 63:13,19 69:22 apologize 73:8 **Aetna** 158:10 102:22 103:3 163:20 149:7 191:2 183:20 277:1 109:12 141:21 affect 94:22 97:10 algorithm 140:7 295:8 156:5 219:11 187:22 altered 334:5 97:21 185:1 203:5 222:15 323:21 327:6 apparent 77:2 appeals 350:12 afternoon 212:9 altogether 126:22 236:3,6,9,10 339:7 340:15.15 267:8 268:16,21 339:10 Analyst 2:15 **appear** 64:20 220:21 age 39:18 42:1,5,6 269:7 277:13 **AMA** 36:20 199:12 analytic 18:4 260:15 297:2 278:5 333:1,22 ambulatory 289:22 **analyzed** 335:20 appearance 199:3 42:9 107:13 anatomical 102:14 130:20,21 132:22 335:11 340:3 290:13 293:5 **appears** 105:12 140:11 168:19 ALISON 2:4 294:18 303:16 126:21 188:20 236:21 and/or 295:2 applicable 55:4 188:14 199:9 alive 36:13 304:1 307:16 200:4 **all-cause** 4:6,9,18 308:21 318:1 **ANDREW** 2:12 application 27:19 agencies 145:3 5:1 13:16,19 14:7 322:5 **anemia** 217:17 32:10 51:6 162:20 agenda 12:22 13:10 15:7,22 59:4 96:5 America 1:18 anesthesia 291:7 185:13 195:5 99:3 107:10 **American** 2:5 3:5 13:12 292:16 307:12 216:3 251:3 aggregate 142:5 128:20 156:12 4:17 220:22 221:6 anesthesiologist 267:15 279:8 aggregates 230:5 166:17 195:13 241:11 293:16,17 309:12 applications 10:1 **ago** 175:6 217:2 266:13 288:7 AMI 4:19 15:8 angioplasty 232:20 112:3,9 182:13 angst 170:13 222:6 228:22 327:1 334:16,17 40:21 267:5,20 215:1.3 346:2,3 348:10,15 230:15 268:5 268:9 269:18,20 176:11 **applied** 12:1 188:6 271:12 322:3 348:19 349:3 270:5 281:20 **Ann** 1:14 193:14 **apply** 143:17 agree 21:2 22:8 all-condition 4:11 299:14 194:18 197:1 163:14 176:5 23:15 35:11 43:10 AMIN 2:10 6:3 198:3 328:6 191:8 199:22 131:8,12,15 44:21 54:1 68:17 135:11 143:4 13:3 15:10 96:9 **ANNE** 2:12 appreciate 6:18 148:3 190:9 192:6 93:4 105:15 147:5 103:13.17 164:4 annual 208:3.17 7:11 61:19,21 162:16 172:12 196:4 198:20 167:1 178:16 226:21 113:20 264:5 182:19 188:9 204:10 179:3 185:17 annually 226:5 352:11 213:4 224:10 all-payer 60:19 214:4 216:13 answer 20:16 50:2 approach 37:10 59:6 85:4,4 254:21 255:16 61:1 132:17 158:9 220:3,7 265:21 63:10 65:11 73:4 297:9 315:9 159:21 162:8 266:7 326:11.15 117:17 155:19,20 107:12,16,18 agreed 224:9 164:8 165:19,21 326:21 328:14,17 167:1 270:8 147:14 148:8,18

Neal R. Gross and Co., Inc. (202) 234-4433

329:4,9 330:10

344:1,6 347:11

350:17 352:1,8

**amount** 28:10

90:20 125:4

170:12 180:7

239:14,21 336:6

60:5 79:2,4 227:3

agreement 26:20

ahead 28:7 32:12

33:15 41:8 45:2

329:19

166:1 167:8

178:17 179:5,19

186:19,22 188:12

179:21 184:20

188:17 189:5

193:2 276:12

Page 354

149:1 172:18

217:9 218:4,12

230:19 238:22

302:20 309:4

228:10,16 229:3

239:1.5.13 267:10

276:10,17 278:12

332:15 347:12

**ANTHONY** 1:20

anticipate 133:18

159:6 187:2

anticipated 34:1

approaches 9:10	<b>ASC</b> 293:5	98:17 190:15	61:10 74:21 77:15	117:2,10,11,14
147:15 192:7	ascribed 60:10	assuming 83:21	77:15,16 79:7	127:7 132:7
228:2 255:14	61:11	148:20 224:5	243:17 282:10	135:11 144:19
336:2	ASCs 309:18	307:2	342:12	162:17 163:6
appropriate 51:18	<b>aside</b> 89:10	assumption 72:5	aversion 49:20 50:4	168:6 180:2
63:12 80:3 117:12	asked 327:16	79:17	50:10,13,18 248:5	183:10 185:5
122:16 132:3	328:12,15 331:11	asthma 177:18	aware 202:20	188:10 199:8
165:22 194:5	334:21 347:9	attaching 210:6	299:12 328:2	205:7 206:16
195:8 203:13	asking 43:6 84:3	attention 8:1	332:21 344:13	226:20 231:16
208:6 216:3	140:17 281:12	attributable 327:10	345:9	239:3,4 240:17,21
284:20 313:17	aspect 24:19 236:4	attributed 38:1	awkward 261:1	241:5,10 263:4,15
319:10	252:8 284:20	314:19 315:8		280:7 283:11
appropriately	aspects 38:8 40:18	attributing 306:16	<u> </u>	290:17 301:16
238:14 319:7	102:19 182:19	322:13	<b>back</b> 6:4 27:3 48:9	315:19 332:18
approval 223:12	229:1	attribution 261:4	48:19 72:20 75:8	334:4
294:13,15	aspiration 292:15	306:4	90:21 91:1 121:4	<b>baseline</b> 305:13
<b>approve</b> 184:21	assemble 9:6	attuned 262:12	123:12,22 124:6	basically 25:18
approved 156:18	assembled 144:3	audience 220:12	125:4,13 129:10	26:20 111:19
176:17 186:14	assess 111:10,15	287:21	152:1 168:17	117:7 222:16
211:1 275:10	134:1 138:20	audit 34:15,20,22	215:19 220:19	250:19
304:1	146:22 163:16	34:22 35:14 51:7	234:16 252:19	<b>basis</b> 118:19 119:10
<b>approving</b> 43:15,17	171:2 172:13	51:9	253:8 260:14	229:18 248:6
43:17 124:9	205:12	audited 34:17	265:2,9 266:6,17	259:10
approximately	assessed 53:4,17	226:20	269:10 279:9	<b>Baylor</b> 1:16
25:19 65:2,4	60:2,19 132:10	AUGER 1:12 111:9	292:12 298:7	<b>Bear</b> 279:6
332:13	assessment 53:2,14	134:6 135:2	306:4 308:6	bearing 331:8
<b>area</b> 70:8 76:20,21	68:6 94:21 141:3	137:15 138:18	317:18 318:4,6	beating 330:6
93:9 120:20 151:1	assiduously 12:6	139:20 157:9	326:19 327:18	<b>beer</b> 240:1
175:20 176:5	assigned 130:11	170:8 182:18	329:22 331:11	beginning 100:21
193:22 336:16	assignment 334:5	189:4 195:10	343:20 350:16	101:8 323:7
<b>areas</b> 180:17	<b>assist</b> 28:18 29:2,10	197:10 201:8	<b>background</b> 17:10	begins 21:11 22:16
334:22 335:15,20	29:14 30:7,10,19	203:11 207:16	24:18 171:11	24:1 33:7 43:22
351:17	31:2,7,11,16,18	322:1	backwards 126:12	47:2 56:6,21
<b>argue</b> 271:17	32:5,6 33:18 34:9	August 350:8	<b>bad</b> 303:11 317:2 322:7	62:17 66:6,20
280:20 319:13	34:10 35:4,7	<b>available</b> 11:6 46:6		71:9 86:6,20 95:5
322:17 341:14	Assistant 2:10	68:16 69:13 91:4	<b>balance</b> 255:19	96:2 106:6 107:1
argument 168:13	associate 211:1	110:13 130:10	<b>balloon</b> 30:16	108:5 110:5,19
arguments 283:4	associated 30:6,9	131:6 133:21	232:22	128:6,16 134:19
<b>arm</b> 298:21	31:15 49:15	135:8 145:2 147:3	ballpark 64:19 barrier 237:7	137:9 138:11
<b>arm's</b> 45:21 46:10	119:14 174:9	159:17 172:17		139:13 157:3
art 228:14	176:11 202:18	213:13 229:9	<b>barriers</b> 45:5 <b>base</b> 42:21 164:1	169:22 190:6
<b>arterial</b> 101:13	254:1 268:20	231:17,20 232:16	<b>based</b> 19:7,9,20	194:14 196:18
artery 4:4,7 18:18	318:17	233:12 245:1	42:10 47:19 48:1	197:18 202:7
18:22 20:1 30:4	Association 2:6	259:14 278:12	69:8 74:8 78:19	206:5 207:3 209:3
30:12 31:3,9 54:4	212:13 241:12	318:8 327:12	94:14 110:12	224:17 226:11
54:14 57:2 96:7	293:17	average 14:22	114:5 116:16	271:4 272:2,16
102:3 348:12,13	<b>assume</b> 38:21 82:15	48:16 55:8,10	114.3 110.10	279:22 298:14

,			1	
behalf 256:19	247:16 259:15	<b>board</b> 7:20 215:17	brought 145:9	301:2,8 302:7
<b>behave</b> 115:3	261:19 276:12	260:18 329:20	273:11 279:8	CABG 4:5,7 15:5
believable 253:11	277:8 282:9 299:9	335:8 338:3	333:9 334:11	20:11 28:15,18
<b>believe</b> 67:13 89:4	300:21 301:3	350:10	Bruce 1:9,11 6:14	29:9,12 31:15
120:13 133:20	302:9,18 315:22	Bob 99:10 109:11	20:18,20 33:12	32:3 36:7,9 42:3
155:13 174:16	323:4 325:10	112:11 113:19	35:20 52:16 95:15	44:11 50:7,13
188:13 222:6	beyond 62:22	119:21 121:8	97:13 104:7 183:9	52:19 58:22 59:5
233:16 244:19	146:17 183:11	border 151:4	183:20 251:20	65:1,3 69:6 70:1
262:21 266:8	261:17	Boston 129:19	328:6 346:6 351:8	74:16 79:22 91:7
268:4 299:1	<b>bias</b> 315:7	140:8 191:1	Bruce's 103:12,19	93:8 273:11 346:7
300:22 331:14	biased 239:4	202:22 205:1	104:1 289:19	CABGs 53:8 64:19
believed 263:8	<b>big</b> 38:17 46:17	212:21 293:2	buckets 126:21	75:5
<b>belt</b> 10:17	114:22 149:18	Boston's 2:22	127:6,15 244:2,16	calculate 68:12
benchmark 301:19	154:4 177:8 218:9	<b>bowel</b> 291:6 305:2	247:13 254:18	160:7
benchmarking	235:7 268:12	brain 338:11	<b>build</b> 158:13,14	calculated 126:2
245:21 246:9	269:21 281:13	brand 152:21	building 281:18	282:18
<b>Bender</b> 293:15	<b>bigger</b> 183:9	break 129:6 220:10	built 282:19 286:4	calculating 133:6
beneficiaries 14:12	283:18 299:15	brief 20:13 129:15	Bulger 1:15 62:10	283:11
14:16 130:20	<b>biggest</b> 343:5	191:6 204:7 221:3	66:12 71:15 87:4	calculation 244:17
Bernheim 2:17	<b>billing</b> 132:10	briefly 57:15 99:6	104:7	calendars 349:14
99:13 265:11,12	bipolar 337:1	129:12,14 221:8	<b>bullet</b> 52:6	350:7
266:1,5,8 273:10	<b>birth</b> 199:20	247:21 265:5,10	<b>bunch</b> 12:5 158:4	calibrated 108:22
275:18 276:9	200:18	351:15	bunching 262:10	calibration 133:4
277:14 278:3,17	<b>bit</b> 12:21 70:20	BRIGGS 1:13	<b>burden</b> 23:7 60:2	140:20 192:13
278:20 279:4,7,13	103:1 111:5	126:19 150:16	249:21 256:1	203:15 227:22
281:11 286:8	118:11 145:13	277:11 310:9	burdensome 93:14	calibrations 108:20
288:11 331:2,3	146:11 147:19	Brigham 1:22	<b>burned</b> 206:20	California 42:18
Berry 136:4	148:15 150:6	bring 35:17 215:19	<b>burns</b> 151:14	60:19 164:9 167:8
<b>best</b> 52:11 75:13	155:22 157:14,17	216:10	<b>BURSTIN</b> 2:11	276:4
103:19 104:1	158:21 163:15	bringing 75:8	97:12 125:1	call 16:5 91:16,19
156:1 182:16	169:16 183:12	289:21 314:16	215:11 250:14	97:3 98:18 205:18
260:11 273:20	202:1 210:18	broad 291:4	329:14 330:13	253:5 344:14,15
311:15 339:6,9	231:3 232:4	<b>broader</b> 137:2	Business 2:7	344:16 345:11
346:11,15,18	242:17 245:4	251:20	<b>buy-in</b> 260:12	346:13,17 347:10
347:6	257:16 267:7	broadest 131:16	<b>by-side</b> 346:9	347:15 349:12
best-in-class/com	285:4 288:21	broadly 97:17	<b>bypass</b> 4:4,7 18:18	350:5 352:12
16:6	289:20 310:14	169:6 170:15	18:22 20:1 30:4	<b>called</b> 9:22 30:15
<b>beta</b> 188:3	321:19 345:1	broken 48:21	30:13,14 31:3,9	101:16 333:7,14
<b>better</b> 8:8,9,9 22:5	347:2	bronchiolitis	54:4,14 57:2 96:7	333:15,17
23:13 51:10 53:5	<b>black</b> 10:17	191:12	348:12,14	calling 7:20 252:19
53:15 77:15 85:5	<b>blanket</b> 283:16	bronchopulmona		252:21
85:8 88:2 89:7	bleeding 292:14	219:12	$\frac{C}{C}$	calls 15:17 126:18
124:11 171:16	311:3,3,4,19	Brooks 1:14 193:16	C 71:16 84:11	314:18 332:14
173:16 174:8	318:17,18,19	194:19 197:2	108:17 140:19	350:21 351:7
182:10 187:3,11	319:15	198:4 202:16	189:5 203:16	campaign 254:8
187:15 200:5	blessing 283:16	206:12 207:10	227:19 274:19	cancer 150:19
205:10 235:6	<b>bloating</b> 292:14	209:9 328:6	275:1,15 300:16	151:10 336:9
	1	I	I	I

		1	1	
Candidate 4:3	careful 169:2	320:13	46:17 50:3 51:1	248:18 334:2,8
<b>capture</b> 73:5 74:2	carefully 11:7	<b>Cath</b> 235:20	58:18 80:19 84:14	changes 39:9
151:8 210:1	269:1	catheter-related	89:20 98:3 115:16	235:14 242:13
290:18 308:3	<b>Carol</b> 2:2 8:5	101:20	120:3 130:18	248:20 267:14,22
310:13 311:22	carried 250:11	catheterizations	134:7,10 137:17	275:9 281:3
324:8 338:13	<b>carry</b> 7:2 238:18	101:13	138:3 145:15	332:20 333:8,13
captured 60:1	<b>case</b> 23:4 38:2	CathPCI 222:22	154:1 156:8 157:9	333:16 334:1,3
233:13 237:9	48:13 64:21 65:5	235:15 240:22	158:6 159:6 165:5	changing 114:11
captures 291:5	79:16 94:18	Cathy 18:10	169:12 170:20	253:1
cardiac 17:19 18:8	132:18,20 146:1	cause 1:3 18:21	176:16 182:19	characteristics
45:12,13 101:17	153:6 166:9	45:5 167:6 306:20	183:1 188:10	102:8,11 114:6
204:16	167:12 168:15	320:4	199:12 215:16	117:12 188:22
cardiodefibrillator	192:10 198:21	causing 116:10	216:5 229:22	302:8
255:6	224:6 237:16	169:15 319:15	237:12 252:5	characterization
cardiogenic 229:20	239:9 247:17	caution 93:1	274:1 283:3	164:5
cardiologist 99:11	248:5,17 269:12	176:12	330:13	characterizes 119:9
99:12 265:18	285:11 306:11	cautions 220:1	cetera 40:22 194:6	charged 98:4
cardiologists	<b>cases</b> 25:22 26:8	cautious 251:12	309:12	Charleston 300:21
100:19	27:12 36:17 47:16	253:4	<b>CF</b> 203:19,21	Charlson 300:22
Cardiology 3:6	47:19 53:2,4,17	<b>CCNS</b> 1:16	chairing 6:14	301:21 302:1
4:17 220:22 221:7	54:3,7,9,11	<b>CCP</b> 274:21	<b>chairs</b> 16:13 264:16	<b>chart</b> 35:6 140:7
cardiothoracic	112:18 126:4,9,15	<b>CCRN</b> 1:16	351:11	141:6 205:8,10
74:9 75:15	138:22 205:9	Cedars-Sinai 2:3	challenge 79:15	226:21 263:14
<b>cards</b> 42:14 62:15	232:1 267:20	<b>cell</b> 142:16	80:4 133:5 142:8	274:22 333:21
66:2 71:6 86:2	282:11	<b>Centeno</b> 1:16 104:8	146:21 152:5	<b>chase</b> 261:12
92:19 235:10	Cassel 2:9 6:11,13	224:10	154:1,19 208:20	checked 269:2
264:10 296:16	11:14,15	<b>center</b> 1:19 2:1,3	217:12 245:5	chemotherapy
care 1:16 6:19	categorical 237:9	3:4,6 4:11,13 7:18	299:18	205:16
13:21 121:15	243:15	18:3,4 57:18	challenges 153:7	<b>CHEN</b> 1:17
122:6 134:3	categories 89:6	130:3,14 175:5	challenging 51:13	<b>chew</b> 98:17
145:17 151:5,18	102:13 107:16	212:21 265:19	82:1 111:10,15	<b>child</b> 163:22 188:20
151:19 154:5,10	234:6 245:8	293:6 294:19	303:14	<b>children</b> 2:22 130:9
154:11,16 178:1,6	258:18 333:2	306:9,12,16	chance 16:21 98:17	130:17 131:17
181:16,17,17	categorization	308:21 315:18	213:18 236:15	153:2 154:5,6,12
182:10,21 193:8	79:21	318:1	338:4	160:16,17,20,22
194:5,22 200:1	categorizations	centered 291:5	Chanch 278:17	161:1,5 168:10,12
202:19 208:18	88:19	<b>centers</b> 29:4 31:1	<b>change</b> 31:21 43:17	181:19 182:9
233:18 289:22	categorize 61:12	55:9 214:1 217:19	48:2 80:6 88:7	183:15 191:17
293:7 305:7	115:6	290:13 304:2	89:21 121:3 183:6	192:19 219:16
307:11,12,13	categorizing	307:16 312:21	201:5 213:6 247:4	children's 1:13
308:10 313:5,10	247:15	<b>central</b> 102:1	247:5,8 250:1	2:21 17:20 129:19
313:15,17,17	category 61:11	CEO 2:9,10 6:11	253:2 254:16,16	132:19 140:8
314:5,6,7 317:6	62:1 77:13 82:10	<b>CEP</b> 1:18	261:1,5 268:12	152:22 153:1
318:2 322:5 342:2	85:15 89:8 114:22	certain 171:15	281:9 284:20	154:13 161:20
cared 148:21	142:15 151:12,13	199:16 217:18	285:17 324:8	191:1 202:22
192:19 317:13	223:16 234:2	308:12 317:1	334:11	205:1 212:13,21
careers 49:9	240:7 242:3 295:1	certainly 32:2	<b>changed</b> 210:20	<b>CHIPRA</b> 214:1

<b>choice</b> 41:13 100:8	277:1 279:18	303:22 314:10	49:2 52:15,17	190:11 193:10
188:9	286:20 323:11	closely 12:9 269:7	53:21 54:6,15	194:10,17 195:9
<b>choose</b> 116:13	<b>clarify</b> 27:3 43:7	closer 57:22 149:5	55:12 56:11,13,15	195:20 196:12,22
132:6 143:18	81:2 90:4 178:19	350:14	57:4,21 58:4,7	197:8,14 198:1
283:10	245:15 273:11	<b>clue</b> 122:12	61:15,22 62:14,21	201:2,20 202:13
choosing 173:19	275:15 328:12	clumps 115:6	63:8 65:10 66:1	204:20 205:17
<b>chose</b> 208:17	clarifying 249:17	cluster 339:2	66:11,16 67:3	206:1,10,19,22
258:10 313:4	256:16 294:17	clustered 75:6	70:7,10 71:5,13	207:8,15 208:10
<b>chosen</b> 124:16	class 168:5 346:11	clustering 102:8	78:11 81:1 82:8,9	208:21 209:7,14
<b>Chris</b> 11:14 12:2	346:15,18 347:6	341:18	83:6 84:22 85:16	209:21 210:8,17
<b>Christine</b> 2:9 6:10	classic 24:14	CMS 2:20 9:9 19:5	86:1,11,16 87:3	210:22 211:3,14
<b>chronic</b> 132:22	217:17	38:17,20,22 39:14	87:20 90:3 92:2	220:9,15,20
140:12 171:21	classification 79:6	39:19 40:20 41:5	93:22 95:11,19	221:13 223:15
175:7 177:21	classified 21:3	41:12,14,15 47:22	97:4 98:19 99:2	224:8,12,19 225:9
188:15 204:4,5,9	classifying 55:6	57:11 58:11 61:5	102:22 103:9,15	225:18 226:7,16
204:17 218:13	<b>clause</b> 286:4	61:18 69:2 79:9	103:18 104:10,14	227:6,15 232:9
219:8,14	<b>clear</b> 37:21 146:14	80:21 82:22 83:4	104:22 105:9,19	234:8,22 235:9
chronically 175:2	172:4 242:19	87:22 89:1 91:13	106:11,19 107:6	237:2,13 238:5
312:16	250:8 276:6	91:14,16 93:2,8	107:22 108:10	239:6,21 240:15
<b>chunk</b> 158:5	313:13 319:17	93:12,20 98:20	109:2,21 110:9,14	241:7 242:2,18
Cincinnati 1:12	340:14	112:13,15,21	111:1,8,16 112:10	243:9 245:10
<b>claim</b> 83:4 84:1	clearly 270:16,19	125:8,16 129:2	113:12 114:20	246:1 247:20
<b>claims</b> 42:21 60:11	271:14	130:12 131:11	115:21 116:18	249:2,8,14 250:5
63:14 65:4 67:11	<b>click</b> 325:20	159:9 167:5	117:16 119:19	252:18 255:20
67:13 82:22 86:14	<b>clinic</b> 313:5	194:21 197:4	120:6 121:6,22	256:15 259:17,21
93:17 132:14	clinical 9:17 19:9	216:5 221:1	123:16 124:21	260:19,21 262:16
133:16 144:4	19:10,10 20:9	222:22 244:19	126:10 127:21	263:11 264:9,15
145:2 152:1,9	21:4 28:11 29:22	247:18 258:22	128:10 129:1,11	264:22 265:3
157:10,13 158:9	34:18 35:8 51:10	265:1 266:6 268:7	134:4,14,22	270:10,22 271:7
159:5,21 162:8	54:2 59:13 71:20	268:10 272:10	135:19 136:20	271:19 272:6,12
163:6,13,20 164:1	82:16 83:22	286:12 287:15	137:13 138:7,15	272:21 273:7
164:9 166:4,11	100:12 107:14	308:17 309:3	139:9,18 141:9,14	274:1,13 275:14
167:7,8 172:16	114:3 199:21	321:21 330:14	145:5 150:13	276:21 277:9,21
173:18 178:17	200:18 226:20	331:17 332:9	152:17 155:9	279:17 280:5,9,17
179:6 205:11	228:7 235:2	340:15	156:9 157:7,22	281:4,22 284:9,11
206:13,17 235:5	240:17,21 263:4,7	<b>Co-Chair</b> 1:11,12	158:22 161:13	285:9,22 286:19
268:16 280:7	302:18	11:13,16 16:18	162:13,19 163:8	287:7,13,19
313:19 321:1	clinical-based	18:10 20:17,21	164:3 165:16	288:12,14,17
<b>claims-based</b> 40:13	59:20	21:6,13,15 22:11	166:12 167:17	289:4,11 294:16
42:8 58:16 76:16	<b>clinically</b> 101:15	22:21 23:3,14,19	169:9 170:4 171:6	294:21 296:13,22
78:6 80:12 94:2 94:16 163:14	114:2 304:12 clinician's 314:7	24:5,8 25:1 27:2	172:6 175:16	297:7,11,17,21
267:11	clinicians 59:22	28:6 32:7,15 33:11,14 35:15,21	176:16 178:14 180:5 181:2	298:10,18 299:21 300:9 302:22
clarification 25:9	close 58:18 59:10	, , ,	180:5 181:2 182:17 183:13	
41:7 60:14 167:19	65:11 74:20 78:6	38:9,13 41:6 42:13 43:2,5,14	182:17 183:13	305:21,22 306:6,8 307:1 309:13
184:16 185:7	80:21 149:17	44:5,7 45:1 46:19	184:5,8,14 185:9	310:8 312:2
204:21 274:4	186:5 274:19	44.3,743.140.19	180.3 187.12,14	313:21 314:13
	100.3 274.17	77.7,2,12 40.4	107.1,2,11,22	515.21 517.15

				Fage 555
316:5 318:11	colleagues 175:4	<b>coming</b> 92:12 95:20	71:14 72:20 78:18	327:3,15,17,21
319:12 320:2,11	210:1,19 289:12	116:11 127:22	82:10 85:13,14,14	328:2,19 329:5,10
320:20 321:3,12	343:21	129:2,3 176:7	94:9 95:13,13,20	329:13,21 331:6
321:22 322:20	<b>collect</b> 321:2	213:22 214:3	98:11,13 104:5,10	337:19 338:3,4
324:10,22 325:14	collected 133:17	252:14 284:16	104:14 106:2	344:13 347:22
325:19 326:6,12	collecting 144:3	commend 122:3	107:8 108:2	350:4,19
326:16,18 328:11	<b>College</b> 3:5 4:17	316:8	109:22 110:15	<b>committee's</b> 97:14
328:15 329:2,6	220:22 221:6	<b>comment</b> 4:15 5:6	111:3 119:20	committees 7:13
330:5,19 339:21	255:3 293:17	14:3 17:3 24:16	128:1,12 137:4	9:5
340:10 341:2,5	colonoscopies	28:8,14 33:17	138:8 139:10	<b>common</b> 23:16
343:8,19 344:2	290:11,19 292:4	45:9 48:22 49:3	140:15 150:14	26:1 131:19
351:14,22	295:22 298:6,22	52:1 54:17 67:21	156:10 158:22	142:13 162:3
<b>co-chairs</b> 1:10	307:21	75:20 78:17 79:12	170:6 175:21	191:16 197:12
351:4	colonoscopist	80:2 82:14 83:14	181:1 203:6,8,10	199:9 239:22
<b>code</b> 236:20 268:20	306:12	92:3 94:8 97:6	206:1 208:22	273:5 277:2
333:4	colonoscopy 4:22	111:5,17 118:11	211:4 212:5	291:12,12
<b>codes</b> 117:3,4 120:1	15:9 288:20	121:11 165:18	216:14,22 217:2	communicated
120:2,4 132:11	290:11,16 291:11	166:13 169:18	220:4,6,8 224:13	330:9,11
140:4 142:8	291:21 295:13	170:17 172:7	225:10 226:8	communication
268:17	303:6 305:1,10	181:5 182:18	227:7 240:6	176:1,14
coding 117:2	309:21 310:22	195:10 201:8	242:22 246:2	<b>community</b> 114:3
<b>coefficient</b> 67:8,19	312:19 313:12	207:20 209:9	249:14 263:12	152:21 153:4,17
68:2 69:16 71:4	319:5 326:4 349:9	210:7 211:16,17	271:1,20 272:13	154:10,20 170:14
107:19 227:2	colonoscopy-spec	211:20 212:3,7	273:8 274:5	174:19 177:17
<b>coefficients</b> 68:21	302:3	216:16,17 228:18	277:10 279:18	191:12 250:20
70:11 187:16	colorectal 305:7	234:17 238:6,11	280:10 285:21	comorbidities
188:4,7 237:11	Columbia 144:5	242:4,17 247:21	286:20 287:11	276:13
cognizant 351:12	combination 34:13	260:22 262:17	296:13,16 297:8	<b>companion</b> 178:12
<b>coherent</b> 101:15	combine 254:7	273:9 294:3	298:11,11 299:22	193:18 320:9
114:2	combined 28:17	297:10 300:21	312:4,6 320:12	comparability
<b>cohort</b> 20:10 58:17	combines 148:6	301:4 305:5 322:2	321:3 324:13,14	97:15 98:13
60:6 65:1 74:5,15	come 57:11 82:13	335:7 344:8	325:1,15 330:8	comparable 22:4
75:5 76:7 91:3	98:2 114:11,14	349:18,20 351:13	343:19 350:3,5	160:8 161:1
114:15 117:2	124:21 125:1	352:6	352:2,3,7,8	<b>compare</b> 25:4 89:3
148:21 181:20	126:11 152:9	commentary 62:22	commissioned	89:17 90:19
236:20 267:5	159:15 170:18	284:17	194:21 197:4	145:12,20 155:7
269:1 334:6	203:19 265:2	commented 286:10	commitment 7:8	159:13 166:9
cohorts 269:3	269:8,10 292:11	commenting	committee 1:3,8	169:3 223:8 257:6
332:1 334:7	307:6 310:11	129:13 286:9	5:4 6:5,12 9:6	258:11 259:1,10
<b>colinear</b> 229:19	331:11 337:10	comments 11:14	22:12 69:14 70:4	302:15 303:5,17
230:10 232:12	comes 21:21 35:18	12:12,18 16:19	73:21 96:11	303:20 304:7
233:4	119:11 123:12,22	21:6 22:11 23:19	109:19 110:1	compared 82:22
collaboration 19:3	210:19 303:22	28:6 33:3 36:8	179:4 201:3	83:3 89:6 105:13
19:5 41:12 58:19	<b>comfort</b> 53:19	41:7 43:3 45:2	209:18 212:15	122:11 153:19
59:10 61:19	comfortable 43:12	46:20 48:9 56:1	219:12,22 220:13	160:12 179:9
268:10	43:13 262:5	56:15,17 61:16	249:18 268:15	200:14 202:1
collaborative 58:11	comforted 53:12	66:2,17 70:8	269:9 279:1 283:1	207:21 219:16
		,		

233:19 263:10	components 51:2	condition-specific	169:2 177:3	255:10	
288:21 301:8	composite 50:16	192:16 269:2,5	209:12 296:14	contingent 96:13	
327:10	316:14 318:9	333:11	316:19	347:3	
<b>compares</b> 147:13	compounds 218:4	conditional 294:15	considerable 180:7	continuation	
comparing 26:18	compressed 70:20	conditions 100:2	consideration 4:3	310:17	
81:5	compromise	106:16 131:19	16:22 93:12,15	continue 49:8 89:1	
comparison 83:10	186:10	132:22 140:12	171:3 201:4	258:22 289:11	
94:10 130:22	concentrated 154:7	151:10 175:7	207:18 208:16	336:2,17 337:15	
306:22	<b>concept</b> 217:3	188:15 200:14,15	252:7	continued 4:3	
comparisons 146:4	conceptual 301:16	204:9,13 218:9	considerations	continues 197:2	
160:2,3 167:4	conceptually	219:9 225:20	186:21 191:8	262:3	
217:21	214:14	229:21	341:13	continuing 250:3	
compelling 62:6	concern 11:22	conducted 172:22	considered 17:1	continuous 234:8	
competent 100:14	38:16,17 87:12,14	175:5 333:21	71:3 82:4 121:21	243:15	
competing 96:14	140:14 156:17	conference 1:8	148:8 228:7 313:9	contractor 112:16	
96:17 98:7,10	158:2 163:1,9	98:18 351:7	340:2,3,5 342:14	contracts 9:10	
264:8 336:12	169:16 184:10	conferences 52:4	considering 83:8	contributes 307:11	
344:17,19	188:1 218:4,6	confidence 51:18	83:10 184:7	<b>control</b> 33:21	
competition 97:11	259:21 285:16,17	81:9 143:15	222:17	259:15	
<b>compile</b> 211:22	305:11	confident 33:19	consisted 132:13	controversial 37:6	
complement 19:17	concerned 94:17	75:15 305:16	consistency 24:20	85:12 173:5	
complementary	162:20 172:3	308:5	27:6 40:19 41:5	controversy 36:22	
19:16 259:4	218:17 219:4	confidential 331:18	255:10 308:9	329:17	
complete 40:9	concerns 56:14	confounded 306:12	consistent 37:10	conundrum 182:12	
152:3	68:1 69:7 82:10	confused 113:13	52:14 140:5	185:18	
completed 203:5	86:17 94:1 95:1	184:17	155:14 200:10	<b>convened</b> 108:13	
completely 69:20	95:12 171:8 181:8	confusion 97:19	230:16 247:18	convenience 114:7	
185:21 232:5	206:16 217:6	169:10	<b>constant</b> 286:14	315:21	
244:18 257:2	240:6 246:2 249:3	<b>cons</b> 59:16 252:8	construct 85:11	convergent 81:7,18	
282:16 292:21	275:13 283:15	consciousness	134:11 202:17	205:18 213:19	
322:12	285:21 296:14	176:4	301:14	converges 205:19	
completeness	299:22 324:14	consensus 13:22	constructed 301:15	conversation 6:17	
144:20	<b>concert</b> 60:8 80:21	14:10,13,16 15:2	consultation 113:7	16:15 96:13	
<b>complex</b> 100:11	concise 12:12,14,21	37:7 210:6	consumer 45:8	185:19 252:16	
175:2,7 177:21	54:17,21	consented 35:7	Consumer-Purch	306:2,18 328:4	
210:15 218:8	concluded 141:22	consequence 49:21	2:4	329:1,3 345:1,17	
complexity 101:2	conclusion 303:18	consequences	consumers 45:18	346:10 347:19	
complicated 173:17	conclusions 7:1	35:19 48:7 74:4	45:19,22 46:2,10	conversations	
337:18	92:13 300:17	216:10 217:14	46:17 48:14 92:14	328:18	
complication	concordant 59:6	247:22 248:4	255:18 258:14	converted 41:3	
295:14,19 320:5	<b>condense</b> 190:16	251:15 270:3	259:3	conveyed 229:4	
complications	condition 107:15	285:7 286:13,16	<b>contact</b> 91:22 314:6	230:12 237:8	
311:6	131:10 148:3	330:2	contest 231:4	copies 70:5	
component 135:12	177:21 192:17	conservative 68:6	<b>context</b> 50:21 54:2	<b>Core</b> 265:13	
245:18 307:10	193:9 204:4,18	78:10 342:21	83:15,19 269:19	corner 265:9	
308:4,8 314:5	218:13 225:21	consider 124:8	270:4 333:9	<b>coronary</b> 4:4,7,17	
319:21 339:19	226:6 319:11	145:20 156:20	contextually	18:18,22 20:1	
	1	1	1	1	
,			I		
---------------------------	--------------------------	----------------------------	---------------------	----------------------	
30:4,12 31:3,8	51:11 154:15	criterion 81:3,17	D	182:8,14 183:10	
54:4,14 57:2 96:7	<b>counts</b> 340:6,7	202:21 204:22	<b>D.C</b> 1:9	184:11 186:22	
222:4,8 264:21	couple 29:22 33:14	205:2,3,20 250:17	<b>danger</b> 94:16	188:6,11,12,17	
348:12,13 349:2	50:1 71:15 166:6	critical 101:6 291:9	data 19:8,10,20	195:2,6,7 198:6	
correct 25:3 38:19	175:5 181:4	309:9	20:2 24:19 25:5	199:16,18 202:22	
39:21 67:15,16	203:14 235:3	cross-fertilization	25:20 27:7,13	205:7,12,19	
73:4 79:1 109:4	251:13 256:18	244:13	28:3 31:14 36:5	206:13,14,17,18	
112:20 113:18	266:11 284:11	crosscutting 55:1	36:17 42:10,19	209:11,15 222:22	
123:17 153:21	307:8,18 312:3	crosstalk 244:13	45:17 46:3,4	226:19,20,21	
168:9 186:16	<b>course</b> 40:11,12	crosswalk 236:18	47:17 48:3 52:14	227:20 228:5,9	
208:13 240:2	137:19 138:4	crude 236:7	60:11,13,20 61:1	229:5 231:5,9	
299:4,5 322:22	147:22 155:4	<b>CSAC</b> 215:17	61:3 63:14 64:4	232:1 235:2,5,21	
341:1	173:4 187:6 328:8	350:9	64:11 67:14 68:15	237:5 240:17,18	
corrected 250:6	covariate 231:5	curious 63:13,18	69:3 71:19,20	240:18,21 245:1,4	
correction 273:16	covariates 107:15	313:3 318:13	72:1,7 78:20 79:5	245:21 259:1,9	
correctly 27:21	188:10 231:21	current 167:21	79:18 80:11 81:11	263:5,7,10 268:5	
185:4	cover 115:15	239:13 240:4	82:5,16 83:1,1,2	273:6,15 275:4,6	
correlated 77:7,10	covered 62:5 158:3	241:13 243:18	83:21 86:14 88:13	275:12,21 276:4	
120:5	161:1 163:20	currently 19:11	88:14,16 89:2	276:12 277:3,5	
correlating 171:15	167:22	43:18 45:16 89:14	91:16 97:21 98:9	280:8 284:1 285:8	
correlation 67:8,19	covers 160:19	90:12 115:18	108:17 110:13	291:15,20 294:8	
68:2,20 69:16	<b>cozy</b> 286:6	143:1 147:10	121:18 123:13	295:21 299:10	
70:11 71:3 107:19	<b>CPHQ</b> 2:3	172:17,22 173:10	125:18 126:1,1,2	300:16 301:10,11	
120:1 227:2	<b>CPT</b> 117:4 120:1,3	241:2 340:18	126:3,7,8 132:12	302:12 309:6	
276:19	create 44:15 69:17	Curtis 2:18 99:12	132:18,18 133:17	310:10,18 313:19	
correlations 25:6	127:15 153:6	118:10,11 127:9	137:21 141:6,6,8	317:17 323:3,13	
correspond 131:12	210:18 231:15	127:10 221:10,11	143:20,22 144:6	323:20,22 332:11	
corresponding	254:6,8 273:14	221:15,15,17	144:11,13,17,17	339:8 341:7 342:1	
228:13	created 199:2	222:1 228:19	144:19 145:1,2,4	342:5 343:15	
corresponds 132:5	268:16	232:13 233:21	145:4 147:3	data-free 146:7	
191:22	creates 217:19	237:6 242:7	148:21 151:2,3	database 18:4	
<b>cost</b> 9:13 93:17	creating 80:5	243:22 245:12	152:3,6,10,13,14	19:10,11 31:17,21	
100:4 137:16	creatinine 263:20	246:10,20 248:2	156:1,16 157:10	33:2 34:17 39:3	
175:11 226:4	creation 159:16	251:22 253:3	158:19,21 159:17	39:18 40:6,7 42:2	
236:16	credibility 92:17	254:20 257:1	159:18,21 160:8,9	46:14 51:9 157:16	
cost-effective 93:13	credible 171:9	258:9 259:20	161:18,19 162:6,8	158:10 164:9	
<b>costs</b> 23:7 137:19	<b>credit</b> 112:16	262:1 263:14	162:12 163:7,13	166:7,18 167:9	
137:20	113:21 236:12	<b>curve</b> 311:12	163:20 164:14,19	179:6	
counsel 2:12	<b>crew</b> 99:1	<b>cut</b> 10:19 109:14	165:10,14,20,21	databases 34:18	
260:13	crisp 12:18	127:22 285:14	166:1,3,4,5,7,11	155:12 158:9	
<b>count</b> 104:2 120:12	<b>Cristie</b> 2:7 104:8	<b>cutoff</b> 310:19 324:9	167:7 169:12,13	193:2	
139:4	256:15 295:2,5	<b>cutoffs</b> 24:11	172:16,16 173:18	date 24:13 120:15	
counted 38:5	296:6 320:22	<b>cuts</b> 124:13	176:7 177:7	130:7 199:19,20	
334:10	321:13 328:6	<b>cycle</b> 9:1 214:8,9	178:18,22 179:12	332:6	
counterpart 83:11	criteria 86:18	269:16 286:18	179:19,20,21,22	<b>dates</b> 244:10	
counting 335:11	101:22 174:10	cycles 251:3 266:19	180:3,4 181:3	350:14	
<b>country</b> 44:11	214:12,21 290:18	<b>cystic</b> 204:2 219:11		<b>Dave</b> 3:3 18:7	

<b>day</b> 4:2 6:4,7 13:9	324:7 328:22	demonstrate 186:3	detection 303:7	139:22 141:11
13:10,12 14:7	329:15 346:13	215:7	320:12	156:16,19 172:6
118:17 119:4	347:13	demonstrated	determine 189:9	190:12 195:21
310:15 311:11	decision-making	26:12 202:17	294:11	211:17 223:21
315:20 317:8	258:18	denominator 41:14	determined 24:13	228:17 241:1
days 13:17,20	decisions 60:7	290:15	27:20 109:4 173:5	242:4 243:11
14:19,21 36:13	100:22 286:6	department 14:17	determining 140:1	247:21 265:1
72:18 113:17	335:7	181:21 290:12	256:13	283:20 285:13,20
130:17 218:1	declare 261:18	294:18 308:19	develop 58:12	289:1 314:4 326:7
226:2 227:5	<b>decline</b> 270:3 286:8	departments	109:13 117:5	327:16 330:14,20
251:14 262:21	declines 269:21	309:19	130:12 131:12	343:21
267:6 290:22	declining 269:18	departure 274:3	132:13 146:6	developing 11:3
305:15 310:11,11	decrease 276:7	dependent 138:21	180:20 212:19	159:20 192:8
310:14,15,16,16	decreased 333:18	depending 21:18	254:14	212:22 253:17
311:1,2,8,13	decreasing 269:19	77:4 92:11 187:6	developed 18:17	development 25:20
313:6 316:20	<b>deemed</b> 36:18	215:18	19:2 20:3 30:2	27:13 28:3,12
331:7 345:5	256:7	depends 8:17,18	31:13,14,22 40:19	37:19 59:7,11
349:22 351:1,5	default 262:4	40:3 105:22 113:6	41:1 42:9,19	74:10 100:7
de-identify 81:22	defer 44:19 88:20	163:4,15 164:2	59:10 60:17 61:2	111:18 138:19
deal 42:22 51:15	163:2	<b>Desai</b> 2:18 265:17	84:18 101:7 117:2	166:3 180:4 182:2
52:2 141:20 151:9	define 83:19 163:4	265:17	127:8 130:8	223:11 268:13
151:14 176:10	268:17	describe 89:5	132:20 147:8	275:20 294:14
190:18	defined 341:22	190:20	167:7 252:3	301:9 312:12
dealing 22:6 52:7	342:10	described 24:12	268:22 275:19	331:5
70:18 174:13	defining 20:9 123:5	275:7	293:13 301:10	deviation 342:11
195:16 304:15	123:15	describing 252:12	323:15	deviations 135:14
death 37:1 39:4,8,9	definite 247:9	310:13	developer 49:17	135:16 196:5,8
39:11	311:4	description 345:8	70:3 84:15 88:12	device 28:18 29:2
deaths 37:9 39:10	definitely 143:21	346:13	99:5 111:20	29:10,14 30:7,10
<b>debate</b> 6:20 8:4	144:10	descriptions	112:16 164:21	30:19 31:2,7,12
49:8 173:7 318:21	definition 32:3	346:21	165:1 178:20	31:16 32:5,6
debates 7:19	36:9,18 60:6	<b>design</b> 171:18	256:4 265:4 322:3	33:19 34:9 35:4,8
<b>December</b> 269:17	150:17 198:21	191:6 312:9 313:2	327:4 328:18,20	<b>devices</b> 31:19 34:10
<b>decent</b> 170:12	205:14 292:4	designation 119:17	330:9	<b>devise</b> 142:10
<b>decide</b> 85:9 93:2	312:15 342:21	151:8	developer's 146:5	diabetes 263:20
325:13	343:16	designed 46:2 50:9	developers 12:13	diagnoses 23:17
decided 114:1	definitions 34:14	267:4	12:20 13:5 14:1	66:15 142:4,13,14
121:3 191:14	35:13 51:4,6,7	desirable 172:12	23:4 24:9 29:15	191:19 332:7
204:11 257:9	<b>degree</b> 63:15 146:3	180:3	37:16 38:10 41:4	diagnosis 140:16
<b>decile</b> 225:1,2	delayed 311:4,20	<b>detail</b> 340:2 346:21	43:7 49:3 52:20	141:19 142:8
<b>decision</b> 8:17 80:20	deliberations 7:1	347:18	57:5,11,12 59:13	200:9 218:7 244:9
94:6 101:19	252:22	detailed 90:16,20	59:15 60:9 63:10	268:20 292:11
109:16 113:22	delivery 134:3	91:2 165:8 205:8	67:21 75:8 83:14	333:4,5
123:5,15 167:15	<b>delved</b> 143:22	217:1 267:14	93:2 96:11 98:20	<b>diagram</b> 78:13,14
215:17 230:15	delving 343:8	details 24:17 27:1	108:13 109:10	dialogue 49:19
258:3,13 259:13	demographics	185:11 231:14	112:5,10 129:2,12	70:15 255:17
260:8,11 307:13	107:14	232:7 273:17	135:2 137:15	<b>dialysis</b> 14:9 42:12
L	1	1	1	1

				1
306:3,7,9 312:15	293:8 305:12,20	discretionary	306:13 327:2	<b>doctor's</b> 317:6
312:21	306:3 312:14	318:22	330:15,17 340:14	<b>doctors</b> 173:12,12
dichotomous 70:18	316:14 317:4	discriminate 81:13	344:18 345:11	document 35:2
<b>die</b> 73:10,15 199:12	334:22 338:12	81:19 84:11 85:5	346:14 347:4	296:8
<b>died</b> 36:19,21 72:18	341:13	85:7 87:15 109:3	discussions 7:19	documentation
differ 145:15	differentiates	111:13 227:19	60:16 69:9 72:16	34:14,15 35:1,14
158:21 177:12	263:1	discriminates	125:2 330:18	51:5
187:6	differentiating	109:5	344:20 346:7	<b>dodgy</b> 210:11
difference 48:15	258:17	discrimination	<b>disease</b> 292:7 305:2	<b>doing</b> 43:13 47:15
85:10 97:20 105:8	differently 150:5	87:7 133:4 192:13	336:22	54:3,9 55:6 67:14
116:1 120:15	277:2 284:6	245:7 274:18	diseases 129:18	84:2 103:1 149:13
121:15 136:18,21	316:13	276:18 282:14	171:22	177:13 233:2,17
145:17 148:2,14	differs 59:2 146:17	300:18	disenfranchised	236:17 237:3,22
149:18 167:21	difficult 111:5	discriminatory	176:9	238:19 259:6
168:1,5 177:8	144:9 145:13	277:8	disentangle 339:4	265:5 284:6 290:6
198:20 225:1,5	163:14 337:18	discuss 41:12,20	disorder 337:1	308:9 312:19
238:19 239:16	difficulties 316:11	97:3 344:16	disparities 137:22	318:2
271:15 335:21	difficulty 183:22	346:17	138:2,2 161:4	dollar 23:6
differences 116:4	<b>dig</b> 82:1	discussant 346:1	194:7,22	dollars 282:20
154:4 155:17	direct 148:17 149:8	347:20	display 89:2,12	domain 92:18
169:8 170:9 194:8	149:12 150:6,7	discussants 20:18	153:5	215:4
224:6 252:9	222:20 242:12	62:2 223:17 345:3	displaying 258:19	<b>door</b> 232:22
341:17 343:6,11	309:18 318:4	346:5	disproportionate	<b>dots</b> 76:14
different 19:7	direction 40:3,10	<b>discussed</b> 52:3 63:6	341:8 342:14,19	<b>doubt</b> 31:12
26:21,21 29:13	82:13	76:8 96:16,22	disqualified 333:6	downloaded
61:9 77:16 82:13	directly 16:15	147:18 193:19	dissemination	332:11
83:18 89:8 92:13	160:22	207:12 209:15	259:7	downside 118:14
92:19 94:12 99:1	director 2:14	219:12 248:3	distinct 60:7 91:10	<b>Dr</b> 6:13 11:14,15
100:16 101:3	221:20 265:12	288:22	distinction 24:11	17:18 18:1,5,15
102:12,13 115:3	directors 335:9	discussing 161:12	distinctions 8:21	21:2 22:8 23:15
115:17 120:4	disadvantage 319:9	288:18	distinguish 82:19	25:2,10 28:8
122:8 126:18	disadvantaging	discussion 6:20	distinguishing 84:7	29:18 32:13 33:16
127:6 142:3	233:18	10:4,6 13:1 15:21	distortions 217:21	33:22 34:4 38:11
143:17 146:12,16	disagree 92:20	16:5,8,10 32:8	distributed 133:11	38:15 39:1,20
148:5,12 149:10	336:6,17 337:15	48:6 50:21 58:9	156:3	40:15 41:15,21
150:11 153:11	disagreement	61:7 62:3 67:5	distribution 123:2	42:17 43:11 44:21
158:4,16 160:15	55:13	97:1,10 98:7	136:3,3 214:17	45:10 47:13 48:8
166:19,20 169:8	disappear 42:15	103:6 129:15	259:18 327:6	49:18 52:5 53:22
171:21 172:2	<b>discharge</b> 13:17,20	134:16 161:8	distributional	54:8,18,20 57:6
186:20 195:1,16	36:14 73:15	177:10 180:10	261:14,21 262:5	57:17 58:1,6,9
205:19 207:21	150:22 174:10,17	190:15 201:4	262:11	59:14 61:17 62:4
217:22 218:22	194:4 199:19	202:15 207:13	District 144:5	62:10 63:2,9 64:1
219:1,2 238:2,9	discharged 36:12	209:8 211:18	<b>Ditto</b> 351:22	64:18,22 65:5,6
246:11 252:1	36:20 268:1	212:15 217:8	diverticulitis 305:3	65:13 66:12 67:6
259:19 263:17	discordant 82:1,2	246:11 255:22	<b>DNP</b> 1:16	67:16,17,22 71:15
264:7 282:6,10	discrepancies 60:6	262:18 263:12	doctor 129:18	73:3 78:16 79:14
283:1,5,10,17	discretion 319:22	295:3 305:4	322:14	81:21 82:12 83:17

84:9 86:13 87:4	225:19 226:18	<b>draw</b> 77:11,17,19	easiest 279:7	112:22 113:15
87:21 88:11,20,22	227:16 228:19	167:3 247:13,13	easily 189:9 206:14	169:5 171:13
89:13 90:1,10	231:2 232:13	drawing 187:16	easy 76:9 165:7	190:17 209:19
91:12,21 92:3	233:15,21 234:16	<b>drift</b> 54:16	182:14 210:21	244:20 261:19
93:7 95:15 97:12	235:1 237:6,19	drilldown 121:17	217:17 335:4	309:15 317:10
99:10 104:12	238:15 239:14	drive 159:16 216:9	eavesdrop 6:16	322:19 323:8
105:3,15 106:14	240:16 241:22	253:2 255:16	<b>ED</b> 291:1 304:10	346:15
107:9 108:12	242:7 243:10,22	driven 119:15	314:12 316:15	ejection 228:6,6
109:7,11 110:11	245:12 246:3,10	driving 127:16,19	317:3 345:5,15	229:7,8,16 232:17
111:4,9 112:11,14	246:13,20 248:2	145:18 318:3	EDMUNDSON	233:2
113:19 115:7,22	249:16 250:14	322:6	1:17 62:4 63:2	elected 258:17
116:3,19 117:1,18	251:22 253:3	drop 270:17 271:13	67:6,17 86:13	<b>elective</b> 47:15
118:10 119:21	254:20 255:21	dropped 144:19	295:4,7 296:2	72:10
120:13 121:8	257:1 258:9	271:14	297:9 298:20	electronic 205:6
122:1 124:3 125:1	259:20 262:1,17	drove 300:12	300:11 314:15	elegant 312:8
125:15 126:19	263:14 265:11,17	drug 131:2	325:3	elements 83:4,21
127:1,9 129:16,20	266:1,5,8 270:13	<b>drum</b> 190:1,1	education 174:18	327:2
130:1 134:6 135:2	271:9 272:8 273:1	dry 5:1 15:21 327:4	176:1,15	elevation 229:20
135:9,20 136:8,10	273:10 274:15	327:5 331:12,16	<b>EF</b> 229:2,12 234:4	elevation 229.20 elevator 8:6
136:14 137:15	275:18 276:2,9,22	331:17 332:9,12	234:4,4,5,6	eleven 47:5,5 128:8
138:18 139:20	277:11,14 278:3	332:19 334:5	234.4,4,5,0 237:10	297:19
	278:17,20 279:4,7	<b>Drye</b> 2:19 289:14	<b>effect</b> 135:12	eligible 37:4 41:22
141:12,16 145:7	, , ,	•		6
146:19 149:7	279:13 280:7,19	289:14,18 299:5	147:18 317:10	42:11 182:10
150:16 151:6	281:11 282:4	302:10,10 304:8,9	<b>effective</b> 10:21	205:14 340:22
153:13,22 155:11	285:2 286:8	307:5 316:21	93:18	342:7,12 343:17
155:19 157:9	287:10,14 288:11	<b>dual</b> 340:22 342:7	effects 149:19	Elixhauser 301:2
158:1 159:2	289:14,16,18	342:12 343:17	188:18 310:12,21	301:22 302:2,4
161:16,22 162:15	290:4 293:15	<b>due</b> 58:18 61:12	efficient 10:7,21	Elizabeth 2:19
163:2,3,10 165:3	294:20 295:4,7	192:15 239:7	12:10	57:14 289:14
165:18 166:2	296:1,2 297:9	<b>Duke</b> 3:2 18:2	<b>efficiently</b> 317:13	302:10 304:9
168:8 170:8 171:7	298:20 299:5	25:11	<b>effort</b> 193:4 237:17	Ellen 212:12 214:5
172:8 176:20	300:11 301:5	<b>dummy</b> 230:3	250:16 254:8	email 91:16 346:21
179:1,7 181:12	302:10 303:1,12	233:22	257:12 259:5	348:6
182:18 186:15	304:8 307:5	dysplasia 219:13	261:14,20 281:13	emergencies
187:20 189:4,7	309:17 310:9,20	<u> </u>	282:12 326:7	233:19,20
190:22 191:2,4	312:3 313:7 314:1	eager 75:16	efforts 125:12	emergency 14:17
193:16 194:19	314:15 316:7,21	0	261:5 281:8,14	181:21 232:12
195:10,22 197:2	318:15 319:16	earlier 16:18 82:14	<b>eight</b> 33:9 102:13	233:4 314:10
197:10 198:4,18	320:7 322:1,21	252:20 263:13	189:20 197:21	315:4
201:8 202:16	323:14 325:3	269:13 273:16	334:22	emergent 229:18
203:11,22 205:4	329:14 330:13,22	331:11	<b>eighteen</b> 66:22 95:8	230:8
205:22 206:12	331:2,4 340:4,12	early 111:17	170:2 196:20	eminently 294:9
207:10,16 208:12	340:17,20 341:1	170:17 171:2	226:14 240:13	emotional 176:5
209:9 212:8	341:21 343:10,14	242:17 267:22	280:15	emphasis 78:9
215:11 221:10,15	344:4	349:19	either 31:8 40:14	empirical 239:19
221:17 222:1	dramatic 247:4	easier 190:18	42:1,4 60:7,13	empirically 311:9
223:19 224:21	248:20	260:17	61:9 63:10 64:6	<b>empty</b> 264:15
		1	1	1

an aguna ag 17(,12	202.12	101.10	194.6 229.22	amanga 40:4 77-0
encourage 176:13	302:12	191:10	184:6 228:22	excuse 49:4 77:9
178:11 193:8	equal 68:11 69:18	evaluating 132:7	270:15 274:3	151:13
216:6 244:11	317:15,16	169:1	278:3 284:1	<b>exhausting</b> 351:1
330:17 350:1	equally 174:20	evaluation 3:7	339:16 347:18	exist 137:22 181:10
encouraged 131:11	177:20	15:20 57:19 208:4	examination	194:7
309:3	equivalent 131:1	208:7 217:5	139:21	exists 194:19
encouraging 247:9	316:20	218:21 265:20	examine 178:12	expand 255:7
ended 302:8 309:7	<b>error</b> 49:15 73:8	326:22 327:15	examining 175:6	expanded 236:5
endorse 9:21	129:5	evening 317:8	252:14	268:5
123:18 209:20,20	<b>errors</b> 60:12	event 131:2 200:22	example 8:11 28:3	expanding 277:12
219:22 329:16	especially 6:18	293:9,9 310:3	28:4 126:13	expect 136:17
<b>endorsed</b> 8:13,15	11:17 51:14 122:6	320:4	144:11 145:3	187:9 235:21
156:14 185:4	151:4 177:6	events 291:5	151:8 162:6	236:2 277:5 308:8
215:12 219:5	181:17 261:7	292:21 311:7,19	168:12 174:7	324:1 349:18
222:5 331:10	327:4	311:22 316:20	181:21 187:1	350:8
340:17	essential 177:2	eventually 168:14	199:8,19 204:15	expectation 286:17
endorsement 16:2	essentially 46:14	everybody 16:20	282:5 293:1	expected 73:17
56:12,20 57:1	84:12 92:6 149:22	249:8 275:16	299:14	88:3 122:14
96:1,4,12 123:12	233:16 312:13	295:15	exceed 143:7	140:21 143:6,8
128:11,15,19	330:3,5,7	evidence 20:19	<b>excellence</b> 3:4 4:11	149:2,14 248:17
190:2,5,8 209:8	established 11:7	21:1,1,8,10,12	4:13 130:15	248:21 281:2
211:5,7,10 214:7	62:9	52:12 62:1,5,11	212:21 214:1	323:4,5 347:19
216:2 222:1,13	estimate 25:13	62:15,17,20 81:19	excellent 29:20,20	experience 165:12
262:18 264:12,18	27:11 77:8 85:19	104:5,12,18,21	38:6 64:2 99:9	213:3 214:10
269:11 287:9,12	89:17,21 126:14	105:4 134:7,8,16	129:6	330:1
287:18,22 288:3,6	241:12 273:20	134:18,21 174:14	excess 143:12	experimental
325:17 326:1	299:10 308:14	193:15 194:12,14	excited 269:15	171:18
327:13 328:22	estimated 69:15	194:16 223:17,20	290:1	experimented
333:10 334:21	281:10	223:22 224:2,7,13	exclude 72:3,8,9	142:10
345:20 350:10	estimates 77:7 78:2	224:16 225:7	152:11 200:8,16	expert 2:2 100:13
endovascular	80:8 85:18 88:15	251:4,7 253:17	336:9 337:11,12	100:13 103:4
102:18 119:5	89:14 201:21	254:7,14 270:12	excluded 32:5	108:13 140:4
<b>engage</b> 58:12	202:2 239:3	270:15 271:3,6,18	37:12 73:12	181:3 294:1
engaged 100:14	estimating 261:8	295:1,8,9,18	150:18 151:17	300:12 311:18
engaging 255:17	306:15	296:3,5,11,16,18	268:1 311:21	expertise 7:7 10:15
enhanced 20:8	estimation 49:16	296:21 330:2	excludes 120:9	11:1 28:11,12
enhancement 237:4	estimators 55:5	evolve 253:14	<b>excluding</b> 34:9 37:9	experts 59:13
<b>ensure</b> 255:10	et 40:22 194:5	255:14	72:15,17 199:11	100:19 138:19
ensuring 292:7	309:12	evolved 252:6	exclusion 146:17	181:4 351:16
entertaining	evaluate 9:20 13:13	exacerbations	199:5 202:15	explain 146:11
350:22	16:9 49:11,12	203:19	290:18	180:15 332:14
entire 75:5 148:21	147:4 156:6	exact 119:2 198:5	exclusions 36:15	explained 25:14
163:12 166:21	159:12 179:12	233:9 241:5	37:13 72:13	26:3 273:16
215:13	193:7	335:17 350:14	198:14,16,19	explaining 155:5
entity 115:14	evaluated 147:7	exactly 21:18 71:1	199:3,9,14,17,21	explanation 176:6
294:17	174:6 205:9	109:8 120:2	201:5 275:7	explanations 53:11
environment 254:6	evaluates 131:9	145:14,21 150:11	exclusively 185:10	explicitly 286:12

	Ì	Ì	Ì	I
exploratory 141:21	119:15 137:1	336:6	286:6 317:2,17	findings 166:10,10
express 209:22	147:1 160:15	fascinating 65:15	322:22 323:6	330:21 341:19
expressed 108:14	161:4 169:7 176:7	fashion 84:18	feeling 23:12 75:11	fine 87:18 208:9
extended 310:2	201:22 208:18	127:18 264:4	109:12 116:5	285:3
extent 115:1 233:17	215:21 239:12	faster 193:12 253:2	117:8	<b>finesse</b> 74:13
externally 303:9	242:11 251:9	fault 146:5	<b>fees</b> 45:4	fingertips 278:13
extracts 183:11	258:10 262:12	favor 80:12 177:14	fell 236:8	<b>finish</b> 293:12
extrapolatability	284:14 302:4	FCCP 2:21	felt 35:8 41:18 47:9	finished 129:4
181:6	303:18 315:15	feasibility 44:6,8	75:15 80:16 89:18	first 14:18,21 16:9
extrapolate 166:21	321:17 347:4	44:19 47:1,5	89:18 144:12	20:20 23:1 24:6
291:19	factor 121:12,21	86:11,13,19 87:1	147:8 174:19	29:19 30:3 37:13
extrapolating	125:6 188:21	110:10,16,18,21	176:8 179:22	38:2 82:21 100:8
183:20	228:8 256:13	157:8,20 158:2	200:5,11,22	101:10 107:7
extreme 343:3	273:17 318:18	161:17 162:14,17	204:16 210:2	138:5,17 150:17
extremely 12:10	factors 161:9	162:19 163:1	273:2 302:19	155:15 170:7
31:4 50:11 77:3	179:16 213:7	164:5,20 167:16	304:21 313:10	180:9 181:15
293:19 336:11	218:18 231:17	169:11,17,18,21	319:3	199:22 220:13
337:18	305:19	170:2 206:11,12	fewer 174:2 233:19	229:2,6 234:15
	facts 29:22	206:20 207:2,5	276:15	235:20 251:19
F	FAHA 1:21	240:15 242:3,22	fibrosis 204:2	252:11 254:22
<b>FACC</b> 1:21 2:21	failed 30:17	243:3,6 280:6,12	219:11	266:17 267:2,17
face 59:21 72:11	failure 29:4 31:1	280:14 320:20	field 24:19 33:18	268:22 269:17
108:14 149:9	35:10 42:1,5	321:6,9	35:1 147:10 215:7	270:11 275:19
203:4 263:9	219:14	feasible 86:15	216:7 262:2	289:20 291:11
315:13,13	fair 67:20 68:19	93:20 110:12	fields 1:18 27:7	305:15 310:15
faced 146:21	189:2 227:3	133:19 157:10	51:5 63:9 64:18	311:7 313:6 314:6
180:17,18	242:18 268:11	159:11 163:5,11	65:5,13 82:12	314:19 316:8
<b>FACEP</b> 1:18	299:2	163:17 164:1,18	83:17 95:15	327:14
<b>faces</b> 221:4	fairly 32:22 40:8	165:6,15,20	106:14 181:12	firstly 313:8
facilities 13:18 14:9	107:16 238:17	206:14,17 280:8	235:21 312:3	FISHBANE 1:18
82:19 84:8 114:7	fairness 260:7	294:9 321:2	314:1 340:12,20	fit 20:14 109:1
241:11,17,17,20	306:21	features 36:6	345:3	<b>fits</b> 233:6
307:15 309:5	faith 250:11	<b>fed</b> 124:6	fifteen 58:7 280:3	fitting 157:17
318:8 324:1,4	fall 77:20 136:5	federal 10:2 203:7	<b>fifth</b> 295:13	<b>five</b> 24:3 107:4
<b>facility</b> 4:21 14:6	200:4 331:16	<b>fee</b> 36:3	figure 112:22	148:4 227:13
114:16 116:9,17	falloff 305:14	fee-for 37:11	<b>File</b> 39:4,8,10,11	234:5 272:19
117:11 307:3,8,11	falls 106:15	fee-for-service	<b>fills</b> 133:20	332:1 338:12
307:22 308:7,15	falsely 236:15	14:12,15 222:9	<b>final</b> 95:20	341:13
315:3 316:3 323:3	<b>familiar</b> 144:3	feed 10:11	finally 51:12 61:4	fixed 188:17
326:2 349:7	262:6	feedback 9:1	102:5 141:4	<b>flag</b> 210:18,21
facility-level	families 2:5 173:11	125:16 245:16,19	200:16	236:15
291:18	family 35:6	332:18 334:4	<b>find</b> 65:14 115:1	<b>flaggage</b> 210:22
<b>FACS</b> 2:21	<b>FAOTA</b> 2:3	feel 11:21 20:6	135:6 136:2	flexibility 60:21
fact 28:10 29:8,9	<b>FAPA</b> 2:5	95:19 97:13	261:17 276:16	flights 12:16
31:6 55:9 61:1	far 39:14 112:6	122:15 165:14	277:19 278:8	<b>flip</b> 279:1
73:12 79:3 94:8	136:5 174:2 180:1	169:4 180:9 210:5	303:14 339:11	floor 1:8 262:8
94:13 97:17	240:3 245:8 309:2	246:7 275:11	<b>finding</b> 10:19	<b>Florida</b> 17:21
			1	1

	220 17 252 12	070 1 1 ( 074 0	104 10 20	
fluctuations 25:16	220:17 352:13	272:1,16 274:8	184:18,20	general 2:12 71:16
<b>focus</b> 16:3 90:18	forget 48:22 341:11	279:22 280:3,13	fundamental 182:4	181:7 276:11
130:6 131:20	form 50:3 112:5	287:2,5 291:9	182:12	295:15 307:20
132:6 172:20	178:21 335:17	295:11 296:21	fundamentally	317:22
173:19 175:13	<b>forms</b> 246:11	297:15 298:14	187:15 314:9	generalizable
223:21 281:12	formula 69:12	300:4 320:16	<b>further</b> 25:9 142:19	239:18
focused 174:5	299:8	321:7 323:4,21,22	161:12 170:6	generalize 188:16
201:6	fortunate 10:22	324:17 338:12	190:2 215:2	generally 121:20
<b>focuses</b> 217:9	293:19	341:12	325:12 330:16	164:7,13 199:1
focusing 191:14	fortunately 156:2	fourteen 33:9	Furthermore	generate 156:16
folks 98:21 231:14	<b>Forum</b> 1:1,8	189:19 249:11	131:18	329:3
344:21	forums 284:14	272:19 287:5	<b>future</b> 38:20 43:7	generated 188:8
follow 15:19 21:16	forward 32:1 65:22	296:21	115:16 124:11	generic 218:12
78:12 98:18 127:9	87:22 177:1	<b>fourth</b> 341:11	127:4 210:4	219:19
178:16 251:22	209:12 213:22	fraction 88:9,10	237:14 312:7	geographic 343:9
260:21 306:9	216:11 220:2	228:6,6 229:7,8		343:11
309:3 346:20	230:21 250:4	229:16 232:17	<u> </u>	geographically
347:17	254:13 258:12,20	233:2 337:9,12,13	game 168:5 170:17	156:2
follow-up 9:1 16:5	258:21 260:16	frailty 264:3,3	gamed 33:20	getting 54:8,16
33:17 97:2 117:18	283:6 340:8	<b>frame</b> 40:17 41:1,3	gaming 34:2,11	126:14 182:10
155:16 185:14	352:12	41:4 124:5 162:22	35:11 48:5 49:21	231:13 259:22
194:5 293:7 313:4	<b>found</b> 67:10 131:18	166:13 215:20	50:20 51:3	298:5 302:21,21
347:14 348:3	133:1,12 142:1,7	223:13 244:10	<b>gap</b> 21:14 22:7,15	308:2
351:6 352:12	142:15 143:3	305:16	22:19 62:21 66:3	<b>GI</b> 311:3,4 318:17
followed 350:10	150:6,10 154:6	framing 165:1,5	66:5,8 105:1,4,14	give 9:12 17:11
following 4:7,9,16	156:1 161:4	Frank 1:13 145:6	105:16 106:3,5,9	49:13 94:11 99:19
4:19 14:1 15:4	173:15 174:7	150:14 277:10	135:1 136:1 137:5	102:20 112:15
18:22 96:6 99:4	175:8 186:18	310:8	137:7,11 194:17	129:14 135:10
113:14 128:21	191:20 192:11,18	frankly 170:22	194:20 195:2	150:2 154:13
131:9 156:13	198:9 200:12	free 46:7,9 322:22	196:14,16,20	196:3 213:17
191:11 222:4	204:18 205:11	frequency 298:8	214:13,15,16,18	250:5 302:6
264:20 266:14	323:4,5 335:21	frequent 52:3	224:20,22 225:8	311:15 322:10
288:8 289:19	four 44:3 47:2	frequently 175:8	225:12,16 271:8	345:8
290:10 291:21	66:20,22 86:6,20	229:2	271:18,22 272:4	given 16:1 33:2
304:3 305:9	89:5 106:6,9	Friday 348:6	296:22 297:14,19	38:18 142:2,15
313:12 334:9	107:1 108:5 110:5	friends 314:10	gaps 105:17	148:9,20 152:4
348:11,20 349:1,4	110:19,21 128:5,9	front 21:16 27:1	gastroenterologists	158:18 160:6
foot 263:15 338:7	128:9 137:8	70:3 73:20 83:9	302:2 317:22	171:20 174:16
for-performance	138:10 139:13	88:15 94:6 99:18	Gastroenterology	187:7 192:1
283:13	157:3 169:22	137:18 180:19	293:16,18	200:13 209:18
for-service 36:4	189:16 196:17	184:6 315:1	gates 181:3	280:21 308:5
foray 254:22	197:18 202:7	frustrations 11:21	Geary 2:20 99:15	316:17 327:14
force 250:16	206:4 207:3 209:3	<b>full</b> 9:9 12:22 61:2	279:2,2,5,10,16	gives 53:2 84:1
254:17 298:1	225:13,16 226:11	74:2 154:9 266:18	Geisinger 1:15	299:9 315:14,14
forefront 7:21	227:10 234:5	273:21 299:10	gender 132:22	318:9
forego 233:1	240:10 243:4	303:15	140:13 188:15	giving 113:21
foregoing 129:8	249:6 253:15	fully 68:7 98:9	228:12	236:11 258:1
0 0 - 2/10			l	

<b>gizmo</b> 56:16	40:10 49:5,6,14	178:5,5 180:10	263:9 267:21	102:4
<b>glad</b> 214:3	55:16,20 59:1	183:15 192:21	GRIGONIS 1:20	<b>gun</b> 232:19 284:15
Glance 1:19 28:8	65:12,22 87:22	194:1 201:10	155:11	<b>guys</b> 16:14 135:8
54:18,20 78:16	88:1 89:11 97:5	203:15 212:8,9,20	ground 12:5,8 62:5	162:21 286:9
92:3 104:12 105:3	98:21 100:9,10,16	213:11 227:21	180:18	346:6
107:9 108:12	101:3,6,10,11	228:1 242:9,14	grounds 231:9	
109:7 110:11	103:5,12,18	245:7 250:11	group 2:6,7 9:7	$\frac{\mathrm{H}}{\mathrm{H}}$
111:4 116:19	112:12 114:10	263:19 264:1	18:16 19:15 21:7	half 328:10
117:18 165:18	118:17 122:3,12	300:18 302:6	28:7 36:1 42:18	Hall 1:9,11,20
223:19 224:21	126:20,21 141:9	315:2 322:13	50:14 72:2 75:14	20:21 21:15 23:3
225:19 226:18	145:6 161:14,15	329:18	87:13 109:22	24:8 27:2 33:14
227:16 231:2	162:15 176:18	goodness 109:1	150:4 151:16	35:21 43:5 44:7
233:15 234:16	177:1 182:7	Gotcha 187:12	156:11 158:7	45:3 47:9 52:17
238:15 239:14	186:21 190:13	gotten 184:17	164:2 167:11	54:6 56:13 57:4
240:16 241:22	201:3 209:16,17	government 7:3	209:22 210:2	58:4,7 61:15,22
243:10 255:21	209:19 210:10,14	272:10 342:15	238:13 243:1	62:14,21 63:8
262:17 276:22	211:20 215:15	<b>grab</b> 220:13	259:19,19 286:2	65:10 66:1,11,16
287:14 328:7	216:1 230:21	graft 4:5,7 18:18,22	288:13,15 292:8	67:3 70:7 71:5,13
<b>go</b> 20:20 28:7 31:1	242:15 252:19	54:14 57:2 96:7	293:14 294:6	78:11 82:9 83:6
31:10 32:12 33:15	253:4 254:3,13	348:12,14	303:4 304:16	84:22 86:1,11,16
41:8 45:2 54:13	257:11,18 261:3	grafting 30:13 31:4	322:4 340:11	87:3,20 90:3 92:2
56:2,18 70:9	268:15 277:4	31:9 54:4	351:16,20	93:22 95:11,19
119:4 124:22	283:9,11,13	grafts 30:5	grouped 102:12	97:4 98:19 102:22
134:16 145:6	289:18 290:2	graph 81:10 85:2	grouping 55:8	104:7 152:19
152:1 153:2	305:22 309:9	136:2	142:9	153:21 167:18
168:17 170:7	317:14 325:8	graphic 89:15	groups 19:4,14	175:19 181:2
175:18 178:2	326:13 331:13	graphical 228:2	170:20 176:9	184:8,10 185:9
184:14 193:16	336:2,16 337:14	graphically 108:21	228:13 238:3	187:14 189:1
218:1 220:1	342:18 343:5,6	graphs 76:3 140:21	271:16	209:21 210:17
223:19 262:14	344:10,20,22	<b>gray</b> 76:20,21	guarantee 254:4	220:9,20 221:13
270:11 283:5	345:10,17,22	great 22:21 27:22	guess 44:9 53:10	223:15 224:8,12
298:3 307:14	349:11,17	28:9 34:6 58:2	57:5 104:16	224:19 225:9,18
308:6 317:2	gold 79:18,21 81:4	66:1 73:3 97:13	122:22 146:9	226:7,16 227:6,15
320:21 322:19	81:5,6 204:22	137:13 141:20	167:2 168:6 176:4	232:9 234:8,22
346:14 349:18	205:7	224:12 257:1	184:10,15 196:1	235:9 237:13
350:9	good 6:3 8:11 22:4	262:1 271:7 276:9	212:9 213:19	238:5 239:6,21
goal 262:14 329:2	27:6,16,20 32:22	277:14 278:20	228:10 238:12	240:15 242:2,18
goals 15:16	35:13,14 37:3,16	281:15 282:5	279:15 313:19	243:9 245:10
<b>goes</b> 6:21 10:5,6	45:7 50:5 51:4,9	284:4 294:2	320:3	246:1 247:20
87:9 92:16 123:19	53:5,15 61:20	303:13 307:5	guidance 122:20	249:2,8,14 250:5
124:9 138:3	68:19 88:2 93:10	310:20 313:7	209:19 284:18	252:18 255:20
215:12 229:10	93:11 105:3	316:21 318:15	<b>guide</b> 176:14	256:15 259:17,21
273:19 314:11	133:15 136:15	323:15 351:20	guided 209:17	260:19 262:16
going 9:12 12:1,3,7	140:20 142:9	greater 36:4 59:21	guideline 59:6	263:11 264:9,15
12:15,17 14:2	144:12 147:8	63:15,20 135:18	guidelines 174:8	264:22 284:11
15:3,3 17:9 21:9	151:22 174:18	139:3 201:13	213:4	285:9 288:14,17
22:22 27:15 29:4	176:6,10 177:22	208:19 222:9	<b>guiding</b> 101:7	289:4,11 294:16

	1	1		
294:21 296:13,22	77:18 95:16	234:18 279:3	101:20	274:7,11 279:21
297:7,11,17,21	299:17	335:13	hesitant 256:11	280:3,12,15 287:1
298:10,18 299:21	Harborview 2:1	heard 32:18 36:8	heterogeneity	287:5 297:14,19
300:9 302:22	hard 12:9 33:21	36:11 67:22 90:9	277:6	298:7,12,13,15,16
305:21 306:6	145:18 170:18	130:1 147:18	<b>Hi</b> 18:1 61:17 88:22	299:19 300:3,7
307:1 309:13	195:14 228:21	181:4,14 221:14	91:12 93:7 99:10	319:8 320:15,18
310:8 312:2	241:4 255:18	250:20 284:13,21	221:10 265:11	321:6,10 324:16
313:21 314:13	312:12 313:18	287:16 299:6	279:2 289:14	324:20 336:11
316:5 318:11	338:11 339:3,16	hearing 11:2 22:12	hierarchical 80:14	high-frequency
319:12 320:2,11	harmonization	110:1 146:13	102:7 107:11,15	325:4
320:20 321:3,12	41:18 98:8 327:20	166:19 169:10	147:16 267:10	high-priority 197:3
321:22 322:20	331:13 334:13	283:20	294:10 335:19	225:21 226:6
324:10,22 325:14	344:17 346:16,18	heart 29:3 30:22	high 22:15,19	high-quality 79:10
325:19 326:6,18	347:6	35:10 92:16	23:21,22 24:2,3	92:5 109:9 243:16
328:7,11,15	harmonize 335:1,6	heart-lung 29:1	26:16,20 27:22	298:5
339:21 340:10	335:9	heavily 304:10	29:7 33:6,9 38:7	high-risk 29:6
341:2 343:8,19	harmonized 58:14	<b>Hebrew</b> 1:17	40:7 43:21 44:3	high-stakes 78:8
344:2 351:14	74:11 84:18 98:9	HEIDENREICH	44:12,17 47:1,5	high-volume 127:6
HAMMERSMITH	230:18 335:1,4,16	1:21 21:2 22:8	47:10 56:4,8 66:5	154:9
2:12	harmonizing	23:15 25:2 32:13	66:8,18,19,22,22	high/good 90:6
Han 2:19,20 18:8	131:22 192:4	33:16 38:15 42:17	71:8,11 79:1,6	higher 26:12 31:10
57:7 61:17,17	harmony 172:18	44:21 47:13 87:21	86:5,9,19 87:1,8	73:17 75:2 82:4
88:21,22,22 90:1	Harvard 3:7 18:8	105:15 171:7	92:21 95:4,8	139:7 143:6
91:12,12,21 93:7	129:21 191:3	235:1 237:19	100:4,4,4 106:5,9	155:18 161:5
93:7 112:14,14	hate 178:3	270:13 271:9	106:18,21,22	168:10,20 186:17
125:15,15	hazard 238:12	272:8 273:1	107:4,4 108:4,8	202:1 204:13
hand 29:3 55:18	HCUP 132:17	274:15 276:2	110:4,7,18,21	269:20 271:14
154:20 180:8	166:7 295:20	280:7,19 287:10	111:13 123:6	322:14 338:9,15
184:1 289:7	323:3,20	320:7	128:4,8 134:12	339:12
handful 174:4	head 181:9	held 269:4	137:7,11 138:9,10	higher-risk 187:17
handle 37:1	headache 113:22	<b>Helen</b> 1:17 2:11	138:13,13 139:12	higher-volume
handled 149:20	heads 293:16	16:17 111:16	139:16 153:3,11	55:9 133:14 139:6
handling 141:8	health 1:15,16,17	314:2 345:21	153:18 156:4	192:19
Hands 103:20	2:6,7 3:1 14:19,21	help 12:11 30:1	157:2,5 169:21	highest 105:12
handy 278:7,22	129:18 134:1	81:16 82:18 85:3	170:2 175:10	225:2
Hang 221:13	200:9,14 205:7	85:7,17 124:2	189:15,19 192:1	highlight 100:9
happen 152:9	216:17 296:12	184:12,12 338:20	196:16,20 197:13	102:5 212:14
285:18 314:6	334:18 345:5,14	helped 184:11	197:16,17,21,21	217:17 235:11
happened 151:4	345:16	251:10	202:6,11 206:4,8	267:16
246:14 247:1	healthcare 1:14 7:9	helpful 11:17 12:2	207:2,6 209:2,5	highlights 99:19
285:7	37:7 53:6 252:6	17:15 76:3 122:21	223:22 225:12,16	102:20 217:6
happening 7:15	healthier 299:19	323:13	226:9,10,13,14	highly 77:6,10 78:3
9:16,19 270:4	Healthwise 1:20	helping 10:18 11:8	227:9,13 240:9,13	100:14 110:11
284:7	healthy 219:16	216:8 220:14	243:3,7 244:21	133:19 164:1
happens 81:19	292:5 316:18	helps 11:20 83:12	249:5,11 254:12	165:6,15 173:5
335:14	hear 17:6,7 58:1	85:4 246:17	256:8 271:17,22	176:5 198:10
happy 20:15 70:4	130:19 216:19,21	hemodialysis	272:4,14,15,18,19	229:21 280:8
<b>II</b> V			, , -,,->	

,			1	1
294:1	74:21 75:7,7,12	hospital's 68:11	142:16 143:5,10	273:4 299:1
hinge 182:20	78:2 82:3 84:20	hospital-based	145:15 148:1,9	<b>ICCs</b> 273:14
hinted 235:22	89:2,9,16 90:11	154:3 309:4	149:4 150:18,20	<b>ICD</b> 236:18 255:5
history 305:2	90:11,19 91:15,22	hospital-level	150:21 151:7,9,10	<b>ICD-9</b> 117:3 120:2
<b>HMOs</b> 40:12	96:5 99:3 102:9	304:5	151:11,14,17,17	ICD-9s 120:5
<b>hoc</b> 269:4	113:14 114:6,16	hospital-specific	152:12,20 153:3,4	<b>idea</b> 118:3 183:15
hold 11:12 32:14	114:17 116:13	126:8 127:19	154:5,8,11,14,15	183:16 205:9
32:15 141:10	118:5 119:3,9	244:12	154:20 155:4,7	302:6
187:18 201:3	124:6 126:7	hospital-wide 5:1	172:4 177:12	ideally 172:12
284:9	128:20 129:19	222:14 268:14	178:3 183:3 187:2	181:15
holiday 349:21	130:16 135:12,13	269:1 327:1	192:20,20 195:16	<b>ideas</b> 90:2
home 14:19,21	142:2 144:14	333:13	197:7 198:10	identical 84:12
31:10 118:9,17	148:20,22 149:22	hospitalization	201:12,14,18	identification
119:4 152:16	151:22 152:21,22	4:20 102:1 131:10	219:1 222:7 223:5	101:4 246:8,16
307:14 345:5,14	153:1,5,9,12,17	133:1 139:21	224:1 233:18	identified 65:1,3
345:15	153:20 167:6	191:11,17 197:12	241:2,13 243:13	76:11 92:20,21
homeless 342:4	170:13 171:15	288:9 349:5	243:20 244:14,20	100:1 217:7
homogenous 53:9	178:2 181:18	hospitalizations	245:6,15,20	225:21 246:7,14
136:16	182:3 183:16,17	14:11 132:15	247:16 253:6,9	282:9 284:5
Honor 172:8	186:9 187:8,18	133:13 139:5,6	254:10 255:1,8	identifiers 144:14
HOPDs 309:19	191:1 196:5,8	187:5 192:2	256:20 257:3,5,7	222:21 242:13
hope 77:1 124:10	199:13 200:13,17	198:22 201:16,19	257:20 258:1,5,7	identifies 122:18
264:5	200:20 202:22	266:14,14 270:18	260:4,17 267:19	222:6,15
<b>hopeful</b> 9:11	212:13,22 222:2 223:8 241:12	345:14	267:21 268:6,7	<b>identify</b> 10:9 34:5 64:5 74:14 89:22
hopefully 12:8,12 12:22 126:6	243:22 244:5,7,7	hospitalized 160:20 268:8	270:15 282:8,13 327:7 331:18,21	94:14 101:9 117:5
236:14 251:10	243.22 244.3,7,7 257:6 258:11,14	hospitals 13:21	332:10 337:3,4,10	143:18 217:10,12
286:1 312:6 335:6	259:1,10,12 261:9	19:12,22 20:6	338:8,14 339:2,13	223:1 240:19
hoping 221:3	261:18 264:19	25:8,19,20,22	341:8 343:1,4,7	245:3 267:9
302:16	267:2 268:9 288:7	26:7 32:21 48:12	host 74:9	312:13 313:18
Hopkins 17:20	288:19 290:10,12	49:6,12,22 51:13	hour 310:1 328:10	<b>identifying</b> 79:10
horizon 11:18	290:22 291:1,20	51:20 52:2 54:12	hours 350:19	177:12 203:12
Horowitz 330:22	295:21 301:18	55:14 60:1,3 61:8	housed 308:14	254:12 319:4
330:22 331:4	304:22 305:9,17	64:6,10,12,17	huge 45:16 238:19	ignited 8:3
340:4,17 341:1,21	306:4,5 308:7,18	70:16,17 76:10,12	267:20 274:3	ignore 154:21
343:10,14 344:4	308:20 309:19	77:12 78:20 79:11	293:20 338:18	230:11
horse 330:6	311:10,22 313:11	80:6 81:14,20,22	hugely 122:5	ignored 12:6 234:7
HORWITZ 3:8	315:17 316:16	82:2 88:2 90:21	hundred 23:5	<b>ill</b> 175:2 200:21
hospital 1:13,19,22	317:19 319:8	91:1,5,6 92:20	hybrid 240:17	304:18 312:16
2:21,22 4:6,9,16	326:3 332:6	93:6,14,21 100:6	hypothesize 148:19	ill-defined 85:11
4:18,21 14:18	333:18 334:19	109:6,8 111:11		illness 208:2
15:1,4 17:20	336:10 337:22	115:13 125:13,17	<u> </u>	illuminating
18:20 26:2 36:20	339:20 341:10,19	127:14 131:3,17	<b>i.e</b> 230:7	293:11
36:22 37:1 48:16	342:8,9,10,13,15	132:15,19 133:8	<b>IBD</b> 305:1	illustrate 219:6
48:17 54:3,5 61:7	342:16,16,18	133:12,14 136:5	ICC 68:17 69:8	imagine 144:8
67:18 68:8,13	348:10,19,22	136:13,17 138:22	70:1 77:4,5	immediate 304:2
69:1 73:11,16	349:3,8	139:2,7,8 141:1	138:20 139:3	309:15 344:12
		I	Ι	1

				_
immediately	263:7,16,22	239:8	increasing 284:17	219:7
120:11 294:17	267:16 271:10	imputing 233:15	increasingly 7:3	infectious 129:17
impact 47:19 98:6	281:17 283:8	239:13	251:5	<b>infer</b> 166:16
218:9 252:2	291:9 292:1,17	in-person 349:16	incredibly 84:18	inference 163:9
277:11 307:3	293:10 295:12,14	inability 61:12	91:8 257:19 260:8	166:15 167:10
impacts 197:5	impossible 76:19	inboxes 348:7	291:12	inflammatory
implantable 255:5	impressed 298:6	incentive 241:19	incremental 47:19	305:1
implement 11:8	impression 248:12	incidence 226:1	247:4	influence 218:15
32:2 93:14,18,20	260:1	include 32:3 36:15	independent 69:20	influenza 191:12
94:19 113:6 165:7	impressive 135:22	74:7 101:12,20	82:17 108:17	information 3:1
193:5	<b>improve</b> 85:18	102:11,18 112:18	independently	24:10 37:17,18
implementation	178:1 187:10	114:3 115:19	16:10	46:6,16 48:11
24:11 80:20 88:7	194:3 203:9 224:5	116:15,20 121:12	index 37:12 131:9	52:21 56:5,10
109:14 113:11	235:18,21 244:15	141:22 144:14	133:1 139:4	63:3,5 69:13 70:5
175:22 176:13	256:10 261:6	152:13 163:21	191:11 198:21	81:8 84:19 89:20
221:22 223:4,12	272:10 281:8	176:14 201:1	335:12 340:3,5,8	90:16,21 91:2,4,8
240:4 242:19,20	284:3 296:10	215:3 222:13,20	Indiana 1:14	92:14 95:5,10
247:12 250:9	improved 173:22	268:5 291:1	indicated 112:6	100:5 111:6
257:17	235:17 236:1	309:15 313:4	173:1 232:10	119:11 123:1,20
implemented 27:15	275:9,10	334:14 336:18	indication 257:4	124:1,5 125:13
50:8 61:6 69:3	improvement	included 25:8,21	indications 197:12	128:5,9 135:21
193:3 214:22	22:10 62:8 91:9	29:16 37:16 64:13	indicative 227:3	152:8 172:5
249:22 251:19	111:22 112:1,7	76:12 91:3 101:9	303:10	173:16 178:20
308:17,18 309:2	121:17 123:9	102:14 113:3	indicator 116:20	179:3 189:16,21
implementing 8:10	124:7,19 125:19	120:14 155:4	204:4 234:10,19	195:21 209:6
133:18 255:5	126:16 127:16	174:6 200:6 204:3	238:22 239:5	214:13 215:10
implications	171:10 175:14	231:6 267:1 324:2	319:5	216:11,18 222:22
148:16	215:1,9 216:9	338:5	indicators 204:18	223:10 229:4,8,16
<b>implied</b> 314:11	236:2,21 255:17	<b>includes</b> 28:17	indices 302:19	230:10,12 233:12
implies 149:15	256:2,12 269:13	131:16 132:21	<b>indirect</b> 39:22	233:22 234:7,11
204:22 261:5	276:7 297:6	309:17	148:17,22 149:9	237:8 239:11
<b>imply</b> 216:2	improvements	including 28:20	149:13 222:21	242:9 243:11,12
importance 214:11	134:2 174:10,21	30:15 34:8 50:15	individual 64:5	244:4,6 245:22
252:7	174:22 193:8	99:15 102:7	68:8 81:22 117:8	246:18 249:6,13
important 6:15	242:4,10,12 270:5	118:22 138:1	304:5 325:11	253:8 254:8
7:19 11:5 29:11	270:21	141:17 161:8,10	individually 59:1	255:11 256:6,10
54:22 58:22 82:6	improving 144:10	222:16 231:10	infarction 4:19	258:19 259:8
93:8 99:21 122:5	270:16	269:6	266:15 288:9	282:15 287:2,6
132:4 133:21	<b>impulse</b> 314:19	inclusion 290:17	349:5	299:3 300:14,19
134:12 146:8	imputation 228:11	inclusions 199:6	infection 4:13 15:8	315:9 318:8
147:6 168:14	228:14,15 230:3	<b>inclusive</b> 26:1,13	191:10 192:3	324:17,21 325:13
169:1,2 174:20	230:14 237:3,22	102:17 162:4	194:9 196:3,11	327:12,17 328:16
193:9,22 208:16	237:22 238:16,17	incongruities 60:10	201:16 208:1	infrastructure
212:16,19 213:8	239:4	increase 177:19	211:11 348:18	259:11,15
213:12 214:2	imputation-based	increased 20:8	infections 191:15	inherent 299:18
215:22 221:2	239:1	315:4 333:13	193:20 195:12	initial 192:1 222:13
228:7 250:22	<b>imputed</b> 228:9,12	increases 218:3	197:11 204:5,14	248:12 257:12

327:13	197:22 202:7,12	135:7 195:17	<b>IPPS</b> 94:3,20 97:9	<b>JACC</b> 252:12
<b>initially</b> 101:16	206:5,9 207:3,7	interrupting 49:4	<b>IQR</b> 112:22	<b>Jacobs</b> 2:21 17:18
159:10	209:3,6 225:13,17	interval 77:7,8	irresistible 97:6	17:19 18:5,15
initiatives 255:8	226:11,15 227:10	89:14,17,21 324:9	isolated 20:11	29:18,19 33:22
inpatient 13:18	227:14 240:10,14	324:11	28:15 32:3 36:7,9	34:4,5 38:11 39:1
15:1 112:19 113:1	243:4,8 249:6,12	intervals 36:3	50:7,13 52:19	41:21 43:11 45:10
113:16 116:2,15	272:1,5,16,20	51:19 143:16	53:8 59:5 65:1,3	49:18,18 53:22,22
116:22 117:3,10	274:8,12 279:22	intervention 4:17	74:16 75:5 79:22	54:8 59:14
117:13 118:2,4	280:4,13,16 287:2	177:22 222:4	issue 23:11 28:21	<b>Jane</b> 2:19 18:8 57:6
119:10 120:18,22	287:6 297:15,20	264:21 349:2	36:10 46:12 47:8	<b>Jeff</b> 2:21 17:18
121:3,5 131:2	298:14,17 300:4,8	interventional	48:4 51:1,12	29:19 34:4 39:21
147:2 166:7	320:16,19 321:7	100:18,18	55:19 74:6 83:19	49:18 52:9 53:22
173:12 230:16	321:11 324:17,21	interventions 30:14	87:8 92:8,9 98:1	54:6
266:22 291:2	insurance 158:4	171:12 172:1	98:11,12 139:8	Jeptha 2:18 99:11
<b>inpatients</b> 114:3,10	161:2,3,10 168:6	174:3 222:8	153:17 157:14	118:10 120:19
154:17	168:15 179:14,15	281:20 296:9	159:4 161:17,18	127:10 221:10,15
<b>input</b> 57:5,9 93:18	188:19	<b>intra</b> 30:15	162:14,17,21	232:9 263:13
100:12 251:18	insured 161:5	intraclass 67:8,19	166:14,19,20	<b>Jo</b> 1:14 193:14
293:14,15,20	168:2,2,11	68:2,20 69:15	169:14,17 170:7	194:18 197:1
302:18 326:7	insurers 160:15	70:11 71:3 107:18	180:13,21 189:12	198:3 328:6
<b>insertion</b> 31:3,16	intake 121:1	227:2	192:14,16 198:15	<b>job</b> 178:5
31:20 35:4	integrity 112:17	intracranial 101:18	207:17 208:3	<b>John</b> 1:15 3:1 62:3
insights 85:2 294:2	intention 29:4	intrinsic 338:21	209:14 210:9,15	63:1 66:11 67:5
inspiring 81:10	interest 45:9	introduce 17:10,16	215:14 232:8	69:10 71:14 87:3
<b>instance</b> 52:22 98:8	161:10 273:12	99:6 129:12 221:7	235:20 236:19	104:7 216:17
164:10	306:14 332:8	221:8 265:5,10,15	238:15,21 252:14	293:15 346:6
institution 27:10	interested 159:9	289:13 326:20	305:5 306:10,22	<b>Johns</b> 17:19
38:2	161:12 283:20	introducing 57:16	307:5 328:2	join 13:5 241:19
institutions 38:4	284:4	57:20	issues 8:6,7 39:1	joined 17:21
90:8 94:15 173:1	interesting 67:10	introduction 41:11	44:9,13 49:20	<b>joining</b> 130:2,4
236:12	83:20 164:4	191:6 294:22	55:15 81:2 98:16	<b>journey</b> 228:20
instructions 315:14	175:20 236:4	introductions	126:11,17 141:7	<b>JOYNT</b> 1:22 48:8
insufficient 22:16	269:12 299:8	17:14	144:20 145:4	122:1 124:3
22:20 24:1,4 33:7	interestingly	invasive 75:1	156:17,21 157:17	135:20 136:10
33:10 43:22 44:4	274:20	investigate 65:8	157:20 180:10	145:7 158:1 163:3
47:2,6 56:5,9 66:6	interim 223:3	281:6	190:17,18 203:12	163:10 176:20
66:10,20 67:2	intermediate 21:4	investigated 64:9	207:12 215:21	246:3,13 249:16
71:9,12 86:6,10	296:15	investigator 130:3	275:12 286:3	282:4 285:2 303:1
86:20 87:2 95:5,9	internal 223:9	265:18	294:4 329:12	316:7 322:21
106:6,10 107:1,5	252:22	invisible 292:22	330:16 336:6	judged 67:20 299:2
108:5,9 110:5,8	internally 8:12	<b>invite</b> 211:16	<b>Isuru</b> 3:3 289:16	judging 167:14
110:19,22 128:5,9	internet 46:7	223:18 288:14	290:2,5 303:12	judgment 238:13
137:8,12 138:11	interpretation	328:3	<b>it'll</b> 278:8	<b>July</b> 349:19,21
138:14 139:13,17	185:12	<b>involved</b> 74:10	<b>items</b> 344:7	350:6
157:3,6 169:22	interpreted 68:19	100:21 101:11	iterations 41:2	<b>jump</b> 21:15 284:15
170:3 189:16,20	interquartile 21:22	319:14	T	<b>June</b> 349:19,19,19
196:17,21 197:18	22:3 105:10,21	involving 100:15	J	<b>jury's</b> 247:6

justified 96:18	180:5 182:17	Kathy 111:8 134:5	17:5 22:4 25:8	L
200:11	183:13 184:5,14	135:1 138:17	26:14 42:17,22	
200.11	186:5 187:12	139:19 157:8	48:17 53:7 64:12	label 88:6
K	189:2,11,22	162:10 170:7	64:18 65:7,18	<b>labeled</b> 90:8 109:8
Kansas 163:19,19	190:11 193:10	182:17 189:3	70:22 75:19 76:9	<b>lack</b> 37:6 213:10
Kaplan 1:10,12	194:10,17 195:9	194:11 195:9	76:19 78:14 79:20	lag 125:3 309:16
11:13,16 16:18	195:20 196:12,22	197:9 201:7	87:7 89:4 90:22	<b>laid</b> 12:5
18:10 20:17 21:6	195:20 196:12,22	207:15 208:13	91:18 94:2,13,19	lands 202:1 215:18
21:13 22:11,21	201:2,20 202:13	211:22 321:22	96:17 97:9,12	Lara 3:5 221:19
23:14,19 24:5	201:2,20 202:13	<b>Kathy's</b> 172:10	109:19 111:11,12	241:9 245:14
25:1 28:6 32:7,15	206:1,10,19,22	keep 12:14 72:22	116:6 117:5 119:7	large 32:22 67:9
33:11 35:15 38:9	207:8,15 208:10	112:17 167:14	122:14 123:11	91:5,10 105:16
38:13 41:6 42:13	208:21 209:7,14	176:22 191:5	125:7,11 127:17	170:20 171:22
43:2,14 44:5 45:1	210:8,22 211:3,14	201:6 202:14	136:6 142:18	191:22 193:21
46:19 47:7,12	220:15 237:2	201.0 202.14 204:7,17 302:1	145:14,21 152:8	197:5,6 198:16
48:4 49:2 52:15	241:7 260:21	330:6 336:15	155:6 159:8 161:7	213:11 229:11
53:21 54:15 55:12	265:3 270:10,22	<b>keeping</b> 216:7	163:18,19 167:4	239:10 298:21
56:11,15 57:21	205.5 270.10,22 271:7,19 272:6,12	Keeping 210.7 Kelly 1:20 328:7	168:18 171:10	310:7
70:10 81:1 82:8	272:21 273:7	key 144:13 194:4	181:14 186:1	<b>largely</b> 263:4
85:16 99:2 103:9	274:1,13 275:14	199:17 267:22	195:14 209:16	268:19
103:15,18 104:10	276:21 277:9,21	292:19	213:21 215:14,22	larger 26:16 80:11
104:14,22 105:9	279:17 280:5,9,17	<b>KHAN</b> 2:13 13:11	219:4 239:19	83:2,22 89:19
105:19 106:11,19	281:4,22 284:9	104:6 211:22	242:2,14 246:3,13	180:1 222:16,17
107:6,22 108:10	285:22 286:19	212:6 216:21	259:17 261:2,6	255:3
109:2,21 110:9,14	287:7,13,19	278:15 289:9	262:8,9,12 266:1	largest 45:11 75:14
111:1,8,16 112:10	288:12 305:22	348:5	270:7,15 273:4,20	Larry 28:7 54:16
113:12 114:20	306:8 326:12,16	<b>kick</b> 12:4 223:18	276:4 280:20	56:16 78:14 92:2
115:21 116:18	328:7 329:2,6	kid 163:18	281:13 282:14,22	104:8,11 105:1
117:16 119:19	330:5,19 341:5	<b>kids</b> 145:17 158:3,7	284:2 285:15	106:12 107:8
120:6 121:6,22	351:22	217:18	286:11 292:2,10	108:11 111:2
123:16 124:21	<b>Karen</b> 1:22 2:14	kind 8:17,21 10:22	302:13 303:2,5,7	116:18 117:16
126:10 127:21	48:7 54:21 121:7	19:18 24:14 26:14	303:17 304:12,14	161:15 164:3
128:10 129:1,11	121:7,10,22	91:7 104:2 105:22	304:15 309:2	165:17 223:18
134:4,14,22	123:20 126:11	122:2 123:2 180:9	311:5 317:16,22	238:14 239:7
135:19 136:20	134:14 135:19	209:18 227:21	318:16,21 319:6	255:20 262:16
137:13 138:7,15	138:7 139:9 141:9	250:9 262:9 286:6	331:7,9 339:17	270:22 271:19
139:9,18 141:9,14	145:6 146:20	306:21 315:3	346:4 347:3,11	272:12 273:7
145:5 150:13	157:22 162:16	337:22 338:11	349:15	275:14 276:21
152:17 155:9	163:2 175:18	347:18	knowing 11:18	280:9 287:13
156:9 157:7,22	176:19 246:2	kinds 32:20 71:2	29:15 173:18	328:6
158:22 161:13	249:15 250:15	126:18 156:21	263:19 284:4	<b>Larry's</b> 94:8,9,17
162:13,19 163:8	282:3 284:12	180:10 213:13	315:2 322:9 336:3	98:11 260:22
164:3 165:16	302:22 316:6	217:18 286:2	342:1	lastly 294:12
166:12 167:17	322:20	knew 31:15 100:10	known 152:21	337:16 lata 57:7
169:9 170:4 171:6	<b>Karen's</b> 160:14	100:11 281:7	176:4	late 57:7
172:6 175:16	180:13 252:20	know 7:15 8:16	kudos 146:5	Laughter 34:3
176:16 178:14	<b>KATHERINE</b> 1:12	9:20 10:16 11:11		54:19 91:20 95:18
			l	

		1	1	
184:4 206:21	53:5 68:13 79:2	57:21 58:4 61:15	longstanding	90:4 105:11
211:2 224:11	107:20 116:11	64:1 84:10 88:12	284:13 285:16	237:15 271:12
260:2 266:4 278:2	120:15 127:5	90:13 273:15	look 19:18 21:19	320:3 323:2
288:16 289:3,8	162:8 223:22	list 212:1 218:13	32:20 40:1 71:1	Lori 2:20 99:15
326:17	307:8 308:15	332:3 333:2,5	79:8 103:19,22	279:2,15
<b>LAURENT</b> 1:19	312:12 317:18	listed 120:18 199:4	112:4 115:3	lose 126:15 289:6
lead 16:14 50:12	334:18,19 336:13	literally 295:18	117:19 121:4	loses 261:16
85:9 174:22	337:22 339:19	literate 342:3	123:1,3 126:3,12	losing 12:16 275:3
223:17 345:2,11	levels 53:16 311:12	literature 174:15	149:9 164:14	<b>lost</b> 78:14
345:22 346:1,5	life 1:17 200:21	202:18 239:2	167:3 178:4 179:9	lot 10:4 11:11 12:9
347:20	253:15 286:18	252:13 311:5	194:2 213:8	15:14 40:20 51:18
leadership 113:7	light 52:20 83:8,15	318:16	214:10 230:21	65:18 90:18
260:12	184:9 213:9 296:4	little 11:22 12:21	241:10 246:15	100:11 122:4
<b>lean</b> 10:15,20	325:6	54:10 70:19 103:1	255:9 267:5,6	124:13 141:7
learn 325:8 351:18	<b>LIJ</b> 1:19	111:4,9 118:11	269:6 275:22	161:7 165:10
learned 143:22	likelihood 29:7	145:13 146:11	292:10 304:14	170:14 174:14
leave 199:11	<b>limb</b> 102:15	148:15 150:5,6	307:22 316:9	176:18 219:3,10
250:15 289:6	limit 216:12	157:14,17 163:15	317:20 325:11	220:1 231:13
325:12 347:1	limitation 44:15	169:16 170:18	329:22 331:19	239:2 251:13
leaves 105:17	47:22 124:13	183:12 202:1	340:8 343:10,20	270:14 275:3
led 250:15 290:2	147:10	218:15 228:15,21	345:9 348:5	282:6,12 294:7
331:4 351:4	limitations 151:2	242:17 245:4	352:12	295:9 300:13
<b>left</b> 220:21	158:19	257:16 266:18	looked 25:7 26:5	302:17 303:3
legal 260:13	limited 155:1	267:7 285:4	55:3 62:12 72:14	305:4,20 307:15
<b>legally</b> 293:6	179:21 180:2	288:21 289:20	108:20 136:1,10	314:17 318:4,20
<b>legs</b> 119:6	185:13 192:14	310:14 321:19	137:1,3 143:14	325:7 332:8
Lein 2:20 57:13	limiting 199:8	344:22 347:2	161:3 198:7	338:14 339:14
61:16,17 88:22	limits 155:6 169:4	<b>live</b> 295:16	199:18 227:5	347:17 351:5
91:12 93:7 112:12	<b>line</b> 77:11,17,20	location 102:14	228:1 239:17	lots 7:18 231:14
112:14 113:21	148:5 215:2 220:6	116:17 317:19	242:1 248:15	love 251:17
125:7,9,14,15	241:22 244:18	log 149:15	262:20 263:3	low 22:15,20 23:22
length 9:13 15:1	245:8 247:14	<b>logic</b> 114:13 228:22	305:13,14 341:7	24:4 33:6,10
45:21 46:10 310:2	254:21	346:13	342:6,9,13,15,22	43:21 44:4 47:2,6
<b>lengthy</b> 217:1	linear 149:15	logistic 149:13	looking 48:13	48:11 49:11 55:13
<b>lens</b> 251:20	<b>lines</b> 18:12,13	long 11:12 47:17	70:16 72:12 78:13	56:4,9 66:5,9,20
<b>Leora</b> 3:8 330:22	76:14 77:2 115:13	151:18 212:18	87:19 92:6 98:4	67:1 68:1 71:8,12
339:21	171:8 261:21	228:22 230:15	103:12 113:16	79:7 86:6,10,20
<b>Leslie</b> 1:20 150:14	278:10,19	232:14 256:5	121:18 123:20	87:2,9 88:4 92:22
152:18 167:17	<b>link</b> 64:14 65:4	268:21 295:16	135:5,20 142:22	95:4,9 106:6,10
175:18 184:9	281:17	310:14 323:2	171:10 174:3	107:1,5 108:5,9
328:7	<b>linkage</b> 39:22	331:7	184:8 193:20	110:5,8,19,22
let's 164:9 169:19	235:18 236:1	long-term 13:21	194:3 214:12,14	111:13 128:4,9
175:18 206:2	312:18	178:7	227:17 233:14	133:9 137:8,12
264:10 277:19	<b>linked</b> 40:1,5	longer 39:12 80:11	248:16 250:2	138:10,14 139:13
296:16 297:11,17	linking 40:4	124:5 137:17	256:12 304:10	139:17 148:10
325:15	lion's 233:9 258:4	234:20 266:19	310:17 311:10	155:2 157:3,6
<b>level</b> 22:9 44:17	<b>Lisa</b> 3:6 57:13,17	longitudinal 250:3	<b>looks</b> 25:3,6 55:19	169:22 170:3

	1	1	1	
189:16,20 196:17	lowest 105:12	217:16	341:7 352:13	15:2,22 16:3
196:21 197:17,22	181:19 225:1	<b>manage</b> 49:22	<b>matters</b> 303:6	17:12,17 19:7,9
202:7,12 206:4,9	LRI 192:12,17	Manager 2:12,13	MAX 141:6,6	19:20,21 20:3,7
207:3,6 209:2,6	204:10	mandate 292:8	143:20,21 144:11	20:19,22 27:15,19
217:15 218:2	lucky 293:22	manipulations	144:17 145:4	28:13,15 29:17
225:13,16 226:11	lumped 115:11	261:2	159:18 160:8	31:13,14,22 32:2
226:15 227:10,14	lumpers 114:21	MAP 9:22 10:5,11	161:18 178:18,22	32:10 37:20,22
229:12 234:4,4	lumping 114:22	11:19 59:8 98:3	179:13,20,22	40:13,18 41:4
240:10,14 243:4,7	115:9	176:17 286:2	198:6	42:7,9,22 43:15
244:21 249:5,12	lunch 212:11	294:13	MAZE 60:8	43:18 50:6,12,18
254:12 272:1,5,16	220:10,13	March 223:5	<b>MBA</b> 1:11,15	51:14 53:15,20
272:20 274:8,12	LVAD 72:1,10	246:22	McNAMARA 3:2	55:19 57:1,13,16
279:22 280:4,13	<b>LVEF</b> 230:4	marching 8:8	99:10,11 109:11	57:20 58:10,16,22
280:15 287:1,5	232:15 237:9	Mari 2:22 129:17	109:12 112:11,12	59:2,3,7,9,11,12
290:16 297:15,20	LVF 229:6,13	135:9 136:14	113:19,20 115:7	59:15,16,19 60:7
298:13,17 300:4,8	234:2	141:16 146:20	116:3 117:1	60:9,17,22 62:12
320:15,19 321:7	LYZENGA 2:12	149:19 153:13	119:21,22 120:13	65:20,21 66:13
321:11 324:16,21		155:20 157:12	121:8,9 127:1	68:5 69:2,6 70:3
327:8 338:22	M	159:3 161:22	<b>MD</b> 1:11,12,17,17	71:18 73:13,19,20
339:1	<b>ma'am</b> 18:13 212:2	165:4 168:9 172:9	1:18,18,19,21,22	74:10,11 75:8,12
low-performing	machine 29:1 30:14	179:1 186:16	2:5,9,17,18,18,19	76:16,18 78:8,19
260:4	30:15	187:20 189:7	2:21,22 3:2,3,3,4	79:2,3,6,9 80:13
low-quality 79:10	<b>Mae</b> 1:16 104:7	191:1 193:11	3:6,8	80:13 81:5 83:8
109:9 243:16	223:17 224:8	195:22 198:18	mean 49:4,5,13	83:13 84:14,15
low-risk 119:16	<b>magic</b> 52:6	201:9 203:22	54:5 63:3 65:15	87:6 88:11 89:12
low-SES 342:19	magnitude 23:10	205:4 208:13	74:20 84:14 98:16	93:18 94:3,6,19
low-volume 49:6	136:21 275:17	mark 3:4 130:3	135:14 136:6	94:22 96:4,10,15
49:22 51:13,19	main 44:8 132:12	183:12 191:4	144:22 149:5	96:16,21 98:15
52:2 54:5,12 55:7	199:21	Master 39:4,8,9,11	164:18 167:2,3,13	99:3,7,18,20,21
55:19 61:6 139:8	maintenance	match 36:16 38:18	183:19 196:6	100:10,17 103:4
153:5,9,9	123:12 214:8,9	64:7,11,14,21	232:13 252:1	103:10,22 104:13
low/bad 90:6	215:20 222:2	65:4 185:21	260:6 261:10	108:12,15 109:10
lower 4:13 15:7	286:15	222:21	281:15 308:19	109:13,18 110:12
59:22 82:4 116:7	<b>major</b> 44:19 47:11	matched 36:5	311:19	111:6,7,17,21
116:10,12 119:13	100:22 101:10	39:18 64:6 76:7	meaning 101:17	112:7,17,18 113:3
130:18 156:3	102:5 118:21	matching 36:5	145:1 148:18	113:9 114:11
187:19 191:9,15	151:12 219:8	38:17,22 39:3,5	205:15 243:16	122:11,15,20
192:2 193:20	237:7	39:14 44:14,16,16	means 54:11	124:2 127:1,8
194:9 195:11,12	<b>majority</b> 39:17	material 24:22	112:22 126:15	128:19 129:15
196:2,11 197:10	49:9 133:13 139:5	materials 75:22	187:17 224:4	131:7,8,12,13,16
198:12 201:15	192:18 201:18	77:19 90:4 248:1	331:17 339:4,17	131:21 132:1,5,13
204:14 207:22	237:8 311:6	331:15 334:14	meant 16:19	132:21 133:9,16
211:11 219:6	344:16	338:5	255:16	133:20 134:12
248:9 254:1	making 132:2	<b>math</b> 73:6	measure 4:11,13	135:11 137:18
262:10 299:13,20	174:18 258:13	matter 2:2 119:16	5:1 8:15 9:4,22	138:19 141:11
322:18 338:9	281:13	122:18 129:8	12:13 13:16,17,20	142:6 143:2,4,18
339:3 348:17	malignancy-relat	220:17 336:3	13:21 14:2,7,10	147:13 148:3,4,4
	I	I	I I	

154:1,3 156:12,14	293:10,13,21	89:16 90:13,17	median 143:9,13	101:16 115:11
156:19 157:16	294:3,6,8,13	92:5 93:3,5,11,15	228:13 234:9	225:22
158:8,17 159:7,22	295:1,10 296:10	96:14,19,21 97:8	<b>Medicaid</b> 132:14	<b>MedPAC's</b> 66:14
160:5 161:20,21	299:13,16 303:19	97:11,15,18 98:5	144:3 145:3 152:1	106:15
162:2,4,7,9 163:5	303:21,22 304:1,6	101:3 105:14	157:12 158:6,9,13	meet 7:13 250:12
163:10 164:11,18	304:20 305:4	108:19 124:4,10	159:5,8 160:7,11	315:13
164:21 165:1,6	306:3 307:7	125:3 130:8,12,13	160:19 161:4,20	meeting 6:5 97:2
167:6,16 169:5	308:17 309:2,7	132:2,6 133:3,22	162:3 163:5,22	337:20 349:11,16
173:19 176:2,11	311:16 313:3	136:22 145:12	166:3,18 167:22	meets 134:10
177:1 180:6,18	314:3 316:1 320:8	146:16,22 147:2,4	176:8 179:10	287:11 296:11
181:15 183:18,21	320:9 323:15,18	147:7,13 159:15	180:2,12 181:7	<b>member</b> 4:15 5:6
184:18 188:2	326:2,22 327:9,13	162:11 166:8	182:8,11 183:10	14:3 211:15,19
190:8,9,21 191:10	327:16,20 328:21	167:4,5 172:19	183:11 184:11,19	322:9 344:7 350:9
191:14 192:5,7,8	329:16 331:5,9,19	174:12 177:17	184:22 185:10,12	members 18:6
193:1,3,7,19	332:2,4,12,15	186:11 191:7	185:13 186:1,4	45:14 103:14
194:21 195:3	333:12,13 334:10	192:6 199:2,11,16	187:4,7 188:11,16	104:2,7 220:12,13
196:3 197:3 198:4	334:16,17 335:22	199:22 200:6,10	188:19 193:2	243:1 327:21
198:6,9,15,17,20	336:10 337:2,2,5	202:3 210:2,12,16	206:17 210:9,13	328:1
200:7 201:11	337:7 339:15	212:20 213:1,8,9	294:8 339:8	<b>memo</b> 334:15
204:10 207:11,12	340:4 345:7	213:11,13,14,18	340:21 341:10	Memphis 2:7
211:10,12 214:10	346:16,19 348:16	213:19,22 214:3,7	342:8,22 343:2,18	mental 200:9,13
214:11,13,15,16	348:18	215:6,19 216:1,8	Medicaid-insured	mention 204:7
214:22 215:12	measured 298:9	216:9 220:21	168:10	mentioned 38:8
221:3,9,18 222:5	measurement 2:11	227:4 235:3,4	Medicaid-only	52:9 59:9,14
222:12,20 223:11	2:13,14,15 3:5	244:19 245:9	186:18,22	72:19 137:16
223:21,21 228:17	4:12,14 7:10,20	247:7,19 248:3	medical 1:19,20 2:1	138:5,19 153:8
235:14 236:22	7:21 8:2,9,10	251:9 262:20	2:3 3:7 18:3	162:11 222:11
240:16,22 242:1	11:22 52:19 59:17	263:2,15 267:5,11	23:17 129:22	228:19 295:11
243:10,15,16	80:11 81:3 84:16	268:3 269:5	191:3 268:2 336:8	308:11 322:3
245:2 246:6,17	92:17 94:11 129:5	276:11 283:16	336:20 337:10	349:12
247:22 249:19	133:7 147:11	287:16,16 299:6	Medicare 14:12,15	mentioning 140:2
252:2,4 253:17	265:13	302:14,15 303:3	23:8 36:3 40:5,11	157:11
255:1 256:3,4	measures 4:3 6:8	303:15,16 308:16	41:22 42:2,10,20	messages 168:7
261:3 262:22	8:14 9:7,21 10:2	309:5 312:7	61:2 64:11,20	messy 143:22
263:1,3 264:19	10:10 12:20 13:13	314:12 321:21	67:11 106:17	met 1:8 24:21
265:2,10 266:10	15:4,5,6 16:2,7,22	332:20 333:11	120:18 121:4	205:14 257:8
266:12,16 267:12	18:15 19:2,6,15	334:12 344:17,20	130:20 160:17	265:14 327:15
267:18,19 268:5,7	19:17 20:14 22:5	345:3,6,19,21	167:7 222:9	338:3
268:14 269:6,8,10	40:20,21 43:1	346:2,8,9,22	270:19 291:15	method 69:10,11
269:14 270:8	46:1,9 48:1 50:7	347:21 348:9	323:16 340:19	77:5 143:1 146:15
273:19 276:5,19	50:15 51:15 52:10	350:8,21 351:7	343:15,16,18	148:22 253:21
277:19 282:6	52:10 55:4,16	measuring 74:4	Medicare-insured	methodologically
283:9,19 284:19	58:13,15 59:3	130:5 145:14,21	168:19	262:22
286:15,18 287:14	61:5 69:1,9 70:2	264:3 292:9	medications 171:22	methodologies
288:6 289:13,21	71:2 72:22 77:14	mechanism 259:7	medicine 1:21 2:1	321:19
290:7,8,9,9,15,21	78:1,6 80:5,16	mechanistically	170:13 314:10	methodology 50:6
291:8 292:17	84:16 87:10 88:20	238:20	MedPAC 100:1	50:9 51:3 85:1
	l I	l	l	1

100:20 108:22	322:12	275:10 277:2	320:15,19 321:6	<b>MPH</b> 1:13,22 2:4
203:8 221:19	misattribution	282:19 283:21	321:10 324:16,20	2:18,20,22 3:4
287:17 314:22	322:16	294:10 300:18,21	moderately 234:4	<b>Msc</b> 1:12
321:20 332:17	miscategorize 78:4	300:22 301:2,3,14	modified 31:18	<b>MSHA</b> 2:2
methods 46:7 143:3	misrepresent 75:11	301:16,20 302:4,7	39:8 222:13	<b>MSN</b> 2:1
219:20	missing 141:5	302:8	modify 230:21	Muldoon 3:1
metric 53:2 124:9	143:20 145:4	model's 274:18	moment 7:6 89:1	216:17,19,22
138:6 146:6	199:17 228:5,9,12	276:8	89:11 267:8	multi-stakeholder
158:14,15 163:14	229:2,4,22 230:4	modeling 80:14	326:20	7:9
163:22 183:3,4	230:6,7 231:4,8	107:11 147:16	money 240:1	multiple 34:19
201:10 207:19,21	231:12,12,17,18	237:15 238:9	monitor 330:3	50:15 99:14
207:22 256:7,14	231:20,21 232:2,3	267:10 335:19	monitored 248:13	100:15 125:2
284:8 314:11	232:5,16 233:7,10	models 48:10 93:10	248:14	171:21 228:15
metrics 146:2,13	233:11,22 234:5,6	148:5,13 149:10	monitoring 286:12	230:14 237:3,21
<b>MHA</b> 2:7 3:1	234:19 237:5,9,10	149:17 150:2	292:7 330:4	238:16 239:3
<b>MHS</b> 2:17 3:2,5,8	275:12	228:8 238:7	month 54:10,10	307:21
<b>MI</b> 229:11,20	missingness 230:11	301:22 302:9	217:2	multiplicative
270:17 272:11	mistaken 323:1	moderate 22:9,15	months 9:6 137:16	149:15
microphones 17:6	misunderstand	22:19 23:12,16,22	morbid 229:21	multiply 239:12
mid 87:8	83:16 205:2	24:4 26:15 33:6,9	morning 6:3 212:9	must-pass 251:7
<b>midpoint</b> 87:11	232:11	43:21 44:4 47:1,6	246:12	mutable 261:17
miigated 50:19	misunderstood	56:4,9 66:5,9,19	mortality 73:10,20	myocardial 4:19
mike 57:22 224:9	323:9	67:1 68:20 71:8	230:17 248:11,16	266:15 288:8
mild 218:14 292:12	mix 132:18,20	71:12 86:5,9,19	248:19,22 269:22	349:5
million 23:5 137:18	135:13 153:10	87:1 95:4,9	302:5 336:11	
226:4 298:6	168:15 192:10	105:18 106:5,10	mortality/morbi	N
mind 38:13 40:20	224:7 248:17	106:22 107:5	50:16	<b>N.W</b> 1:9
41:5 44:9 57:15	mixed 168:7 253:21	108:4,8 110:4,8	motivate 134:2	Nabat 278:17
123:7 167:15	model 7:9 9:4	110:18,22 128:4,8	<b>move</b> 22:22 57:22	NACHRI 132:18
183:12 202:14	20:10 42:19 59:20	137:7,11 138:10	60:16 75:20 77:12	136:13
231:2 261:10	92:12 102:7	138:13 139:12,16	86:2 96:9 182:6	Nakamura 2:22
302:1 339:22	108:16,22 116:20	157:2,6 169:21	183:3 209:12	129:16,17 130:1
<b>mine</b> 351:18	124:15 132:21	170:3 189:15,20	224:13 246:18	135:9,10 136:8,14
minimal 193:4	133:2 135:13	196:16,21 197:17	251:10 252:17	136:15 141:12,16
240:4	140:18 146:11	197:22 202:6,12	264:10 297:11	141:17 146:19,20
minimizing 236:14	149:13,15 150:8	206:4,8 207:2,6	325:15 326:8	151:6 153:13,14
236:14	157:17 158:17	209:2,5 218:14	<b>moved</b> 234:9	153:22 155:19,20
minority 152:7	165:11,19 188:5,5	225:12,16 226:10	252:15 267:18	159:2,3 161:22
MINTON-FOLTZ	188:12 189:6	226:14 227:9,13	334:8	162:1 165:3,4
2:1 120:8 309:14	192:11 203:15,20	240:9,13 243:3,7	moving 8:13 9:2	166:2 168:8,9
310:5 343:12	204:3,19 227:1,18	249:5,12 271:22	32:1 67:3 178:10	172:8,9 179:1,2,7
minute 129:15	227:21,22 229:13	272:4,15,19 273:2	224:19 242:12	186:15,16 187:20
278:9 279:6	230:4,17 231:6,10	274:7,12 279:21	258:12,19,21	187:21 189:7,8
minutes 129:5	231:16 234:6	280:3,12,15 287:1	260:16 282:2	190:22 191:1,4
190:13 233:1	237:11 238:13	287:5 290:16	287:8 298:18	195:22 196:1
310:1 322:3	239:10 242:12	297:14,20 298:13	300:9	198:18,19 203:22
misattributed	264:2 274:18	298:17 300:3,8	<b>MPA</b> 2:2	204:1 205:4,5,22
	I	Í	I	I

208:12,13	158:11 182:6	nineties 40:7	237:20 238:6	nutshell 7:15
name 17:4,9,15,18	195:3 201:6	Ninety-five 36:2	266:17 269:4,9	
57:17 129:14,16	209:11 219:4	Ninety-nine 30:7	279:8 284:14,16	0
141:14 212:12	295:13 300:5	<b>no-pay</b> 343:13	284:17 289:21	<b>O'Brien</b> 3:2 17:22
241:8 299:8	308:12 318:10	non-capture 73:7	303:22,22 329:20	18:1,2 25:10,10
name's 18:1 265:17	323:18 324:5	non-elective 72:10	333:9 334:11	39:20,21 40:15,16
289:16	330:16	non-endorsement	343:20 350:9,18	41:15 52:5,6
names 265:8	needed 8:3 35:9	211:5	NQF's 75:17	<b>object</b> 210:6
278:15	204:8 245:3	non-hierarchical	NQF-endorsed	obligation 298:4
narrowed 271:10	323:17	335:19	46:8 50:15 132:1	<b>obs</b> 309:15
nation 7:11 64:17	needle 251:10	non-issue 46:14	132:5 192:5	observation 118:19
89:18	needs 9:17 80:22	non-overlapping	<b>NSQIP</b> 239:22	119:2,10,14 291:2
national 1:1,8 2:4	210:3 253:14	69:21	nuances 97:20	304:11 316:16
8:3 61:2 65:3	285:2 298:8 335:8	non-physician	number 27:11	observations 29:21
76:22 89:7 131:14	negative 247:22	183:14	32:22 33:1,1	228:5
159:16 160:2	285:6 314:11	non-provider	38:19 39:2 46:22	observed 52:14
247:8 269:18	neglected 57:6	183:14	60:1 64:19 83:3	149:14
291:20 324:3	neither 60:10	Non-random	89:19 91:10,19	observed-to 140:20
333:18	<b>net</b> 342:10	341:18	99:3 103:13	obstacles 44:19
nationally 77:13	neurological	non-VAD 74:22	140:11,12 150:1	47:11
160:8 170:21	204:16	Nope 110:15	171:22 182:9	<b>obstetric</b> 200:1,6
291:22	neutropenia	normal 233:16	191:22 192:1	<b>obvious</b> 171:17
nationwide 324:4	217:16	234:4	193:21 197:5,7	258:14
natural 175:14	never 8:7 38:13	North 1:18	198:16 201:18	<b>obviously</b> 36:4 37:2
nature 45:4 46:13	338:4	note 35:6 36:10	212:4 217:13,14	38:20 73:11
64:3 82:18 84:7	Nevertheless 229:1	135:3 159:19	218:2 222:17,17	198:11 229:10
151:5 162:22	new 9:7 50:13 81:5	220:10 299:11	233:9 238:9	270:8 337:17
308:5 346:17	99:1 111:5,21	328:5	239:10 241:5	350:22
nausea 292:14	156:14,18 157:16	noted 12:5 41:11	244:22 258:8	occasionally 42:18
NCQA 334:15,15	162:10 165:13	60:4,18 61:8	267:13,21 278:21	occasions 28:21
335:5,16 336:14	166:9 179:5	66:13 74:7,12	282:11 283:5	occur 130:17
337:6	180:18 183:4	193:2	289:7 305:11	133:13 310:22
neat 268:10	188:7 192:22	notes 103:19 104:1	311:10 319:19,19	311:7
necessarily 84:20	236:10 290:1	notice 306:1	327:2 328:1	occurred 205:13
85:19,22 116:6,9	295:10 335:12	noting 59:19 170:9	342:19 346:5	occurring 292:5
118:16 144:22	347:22	notion 53:13 85:6	352:7	occurs 199:19
160:1,22 164:20	<b>newborns</b> 200:17	252:18	numbers 27:10,21	310:4
232:16 233:11	newly 194:21 340:5	novel 140:1 302:11	28:2 32:21 39:16	<b>October</b> 350:13
248:7 270:14	nice 211:18	302:12	67:9 83:1 154:8	odd 258:5
308:1 312:20	nicely 58:10	NQF 2:9,9 4:15 5:1	154:12 186:12	off-the-cuff 73:6
necessary 177:8	Nihar 2:18 265:15	5:6 6:11,21 7:16	188:4 196:4	offer 60:15 84:17
neck 102:15 119:6	265:17	8:12 11:4 15:15	217:19 245:2	84:19 91:14
<b>need</b> 7:5 29:10 55:2	nine 56:8 66:8	69:9 73:22 75:9	277:15 298:22	157:19
55:11 78:19 80:11	272:4 333:1	122:10 166:13	numerator 187:13	offered 255:1
100:7,7,11 104:19	<b>nineteen</b> 108:8	210:1,18 211:15	187:15 290:21	office 192:22
125:18 133:21	157:5 194:16	211:19 213:5	nurses 173:11	290:14 317:6
134:19 147:7	271:6	214:3,6 216:2	nursing 14:6	officially 120:22
	I	I	I	1

- <b>f</b> 4 <b>f</b> 4	On Deine 14.4	1.00 12 201 01	15 0 112 10 112 0	
oftentimes 118:17	<b>OnPoint</b> 14:4	160:13 301:21	15:9 112:19 113:2	overestimation
233:1	onsite 226:21	order 10:13 69:3	113:15 114:5,15	186:10
<b>Oh</b> 34:4 273:8	<b>open</b> 6:20 9:3 18:11	82:22 89:22	114:18 116:2,8,9	overlap 68:9 93:6
276:2 314:1	18:14 62:3 67:5	237:16	116:12,14,22	130:21 338:17,18
okay 11:15 15:10	75:8 102:19,21	organization 45:12	117:4,10,13,21	overlapping 77:3
16:17 42:13 43:3	164:16 223:13	45:21 253:5 259:5	118:2,6,14,15,20	overnight 119:4
43:19 44:5 46:20	278:10,19 295:3	260:13	119:2 120:10,22	overview 270:7
56:1,17 66:16	opening 71:14	organizations 45:6	121:2 122:4	overwhelming
67:6 78:11 83:13	300:10 320:21	45:8	173:11 178:6	39:17
86:1 95:11 99:2	321:13	original 234:17	288:19 290:10,12	overwhelmingly
103:9,21 104:4,15	openness 6:19	274:16 337:19	290:19 291:13	294:5
104:22 108:2	operating 30:12	originally 40:22	292:20 294:18	owner 334:15
114:9 127:21	31:1 35:3 120:21	266:16 268:8	304:17 306:16	P
128:13 142:19	operations 51:16	334:20	308:19,20 309:19	
169:9 184:14	operator 18:13	<b>ORT/L</b> 2:2	312:7 316:9	P-R-O-C-E-E-D
187:13 189:11	212:2 216:15	orthopedic 151:11	317:13 325:4	6:1
190:3 196:13	220:5 278:10,19	out-of 151:21	326:3 349:9	<b>p.m</b> 220:19 352:14
205:21 224:12,13	352:1,5	outcome 21:4 37:3	outpatients 114:4	Pace 2:14 111:19
225:10 265:3	operator's 307:4	45:17 46:4 50:3	178:1	121:10,11 122:22
266:7,11 278:4,20	<b>opinion</b> 140:4	55:15 65:7,8	outputs 148:6	package 254:15
279:10,13,16	250:6,13 287:12	114:19 289:22	149:6	page 76:2
285:20 286:7,21	opinions 252:9	291:5,14 296:12	outside 68:18 136:5	pages 277:22 279:1
290:4 295:5,7	opportunities	296:15 299:12,20	152:12,15 185:22	pain 286:1 292:14
300:11 313:21	255:7	303:15 305:6,16	186:3 212:17	<b>pair</b> 76:14
314:1 323:14	opportunity 7:13	307:12 308:1	outstanding 20:22	paired 73:19 77:2
324:12 326:11,21	41:19 61:8 62:8	312:20 322:6	24:9 28:13 215:14	<b>Pam</b> 345:4,10
341:4 344:6 352:5	204:6 224:20	outcomes 3:6 26:21	over-65 61:2	PAMELA 2:2
352:9	255:2 258:6,16	39:13 46:1 57:18	160:16	panel 72:11,16
<b>old</b> 222:10	260:7 297:1,5	63:16 74:3 224:4	overall 20:21 21:17	100:13,14 108:13
<b>older</b> 41:13,14,16	opposed 25:15	224:5 256:3,10,13	38:6 39:16 56:19	131:15 294:1
42:20 167:10	48:20 93:5 101:2	265:19 308:13	56:22 58:3 94:21	300:12 311:18
once 143:22 172:5	118:20 201:5	313:9 315:3 318:2	95:12,22 96:3	panels 140:4
180:17 201:21	205:20 217:10	322:15 323:8	127:2,8 128:14,18	paper 69:10 136:4
216:15 255:9	222:21 232:2	outlier 80:15 90:19	134:11 147:14	210:20 252:11
350:14	237:3 247:4	111:11 246:9	148:10 153:10,18	284:16
one's 111:4 289:5	253:15 263:4	322:7 323:21	181:19 190:4,7	papers 39:6
one-fifth 65:2	277:22 303:11	outliers 76:11 88:6	191:20 211:6,9	paperwork 260:14
one-third 154:16	322:14	88:9,10 89:19,22	214:18,18 226:1	parallel 18:17 19:3
160:19 187:5	<b>opposite</b> 282:16	90:6,7 91:10	235:19,22 236:2,7	59:12
305:8	<b>OPPS</b> 308:20	122:13,19 141:2	236:22 244:3	parameters 35:20
<b>one-year</b> 267:18	<b>opt</b> 258:7 260:4	142:20,22 143:15	249:15 259:19	part 19:4 28:16
ones 92:21 106:1	opted 245:18 257:3	143:18 243:21	264:10,11,18	29:17 42:6 63:19
115:17 145:8	257:4,6,20 258:5	244:17,21 246:17	271:16 274:18	63:22 70:14,15
146:12 176:17	<b>optimal</b> 63:16	254:12 284:6	277:17 288:2,5	78:17,21 99:15
267:16	124:7 214:19	294:11 324:2,8	325:1,16 326:1	130:13 162:9
ongoing 242:20	optimistic 9:12	outline 291:10	331:22 334:8	168:15 177:9
250:9	option 30:18	outpatient 4:21	overestimate 186:7	179:20 185:6
	I	I	I	I

187:22 191:8	102:8,10 116:6,11	292:11 293:8	131:2,5,19 132:3	per-physician
213:17 239:9	116:16 117:8,11	304:13,18,21	133:6,7,10,13	308:3
244:22 268:13	132:21 152:2	305:2,2 308:12	134:3,8 135:3	percent 19:12,13
286:11,17,17	153:10 188:21	309:6 310:6 327:8	137:19,22 138:5	19:22 20:5 21:17
314:8 320:3	229:11 244:8	327:11 332:3	142:8 145:12	21:21 26:4,9,9,10
326:13,21 329:19	263:20 277:6	334:6,7 336:8,10	147:1,10 148:1,10	30:4,7 36:3 37:7
330:15 346:7	291:4 303:11	336:15,18,18,21	153:18,19 154:2	38:21 39:22 40:5
351:20	304:16 305:18	337:14 338:9,15	154:16 158:15,19	40:8,8 60:5 63:4,5
participate 20:2,6	307:14 310:2	339:1,6,8,10,14	162:1 163:10	64:16 73:7,7,16
58:19 63:17 95:16	312:15 315:13	340:19,21 341:10	164:11 166:21	74:18 75:3,5 78:7
257:9 328:3	317:14,18 332:4	341:18 342:7,12	170:13 174:5,6,12	78:7 79:4,11,11
participating 40:3	337:22 339:18	342:19 343:2,16	175:3 181:15	80:3 89:17 90:5,6
40:7 241:11,14,21	patient's 35:10	343:18 345:15	182:20 190:9	90:7 98:14 100:3
243:13 245:16	116:7	<b>Paul</b> 1:21 20:18	191:9 193:9 195:6	105:7,7,8,11,21
258:15	patient-level 309:5	21:16 23:14 25:1	197:6 199:8 200:4	106:16 122:18,19
participation 64:15	patients 20:10,10	27:5 32:12 33:15	200:7 211:11	130:17,19 135:4
particular 38:11	20:11 28:17,20	35:16 38:14 42:16	212:20 213:11	135:15,17 139:4
50:12 52:18 61:11	29:6,12,14,16	44:20 47:12 87:20	218:22 219:7	143:5,12 144:21
75:7 82:3 97:22	31:1,7 32:4 34:8,9	104:8 105:14	348:15,17	154:14 191:21
111:7 149:22	36:12,16,19,21	171:6 234:22	pediatrics 130:7	196:7,10 201:15
152:10 188:7	37:11 38:3 39:17	235:22 237:18	142:14 147:21	203:2 219:15,17
204:2 210:2	50:14 65:2,3,9	270:11 271:8	151:7 159:15	225:2,3,6 226:2
214:11 231:5,15	68:7,11 70:17	272:22 274:14	161:11 166:17	228:5 231:8,9
238:8 241:3 252:8	72:1,10,16,17	280:18 287:9	171:20 174:1,2,20	233:7 236:8,9
263:3 276:17	73:10,15,17 74:5	320:6 346:6	176:6 197:13	241:13,15 244:20
281:20 283:19	74:16,22 76:8	<b>Paula</b> 2:1 120:7	200:3	245:10,11,13
284:19 315:2	77:12 78:5 90:17	309:13 328:6	<b>peer</b> 318:4	246:6 252:21
329:11 351:3	91:2,6 100:7	<b>Paulette</b> 103:11	penalty 282:17,21	271:11 282:19
particularly 45:6	101:5 102:9 114:8	104:9	penetrance 19:11	290:19 293:3
69:5 76:3 98:6	117:8,20 118:7,16	pay 283:12	59:22	295:20 310:7
125:2 146:3 177:2	119:1 121:19	pay-for-perform	penetration 44:10	323:16,18 324:11
307:16 329:20	133:11,12 142:5	143:2 213:15	<b>people</b> 12:16 17:4	324:12 332:10
338:20	142:12 152:3,15	payer 138:2 162:12	18:11 35:18 37:8	333:15,17,19
<b>partner</b> 45:19	153:18,19 164:11	163:20 166:11	61:4 75:19 77:20	337:9 343:1
258:22	173:11 174:5,6	169:6 176:9	80:14 88:4 97:19	percent/90 271:11
Partnership 2:4	175:2,3,7,9 176:8	179:12	97:21 98:14 99:14	percentage 22:2
10:1	177:21 178:6	payers 131:3 169:8	103:21 112:6	26:3 88:6 144:18
pass 350:2	180:2 194:8 197:6	payment 87:11	114:10 115:17	151:21 175:9,10
passed 13:18 14:5,7	199:11,12 200:1,2	281:1 321:18	122:10 167:11	241:1 256:20
14:19,21 348:12	200:8,19 204:13	<b>PCI</b> 4:17 15:8	200:12 210:5	279:14
348:14,16,18,21	219:1,13 222:9	229:9,10,14,17	231:4,7 240:1	percentages 78:5
349:2,6,9	229:15,19,20	241:13 248:19	246:18 273:12,19	<b>percentile</b> 105:5,5
<b>passive</b> 313:1	230:5 231:18	252:5 255:4 PCIs 241:2	277:22 278:16	225:5,6 324:6
patient 28:22 29:7	232:15,19,21	<b>PCIs</b> 241:2	284:5 286:11	percentiles 136:7
29:9 30:11,19 35:2 37:3 41:13	240:19 241:16 267:6 268:1,8	<b>pediatric</b> 3:5 4:11 4:11,13,14 15:6	289:5 292:6 298:3 308:6 314:2 322:4	percutaneous 4:16 74:19 222:4,8
42:11 100:20	270:19 281:21	4:11,13,14 15:6 129:17 130:13,22	308:6 314:2 322:4 332:13 342:2	264:20 349:2
42.11 100.20	270.19 201.21	129.17 130.13,22	332.13 342.2	204.20 349.2

	1	1		
perfect 44:15 92:11	periods 26:22	311:13	176:14 252:22	213:6 237:20
182:1 234:13	80:11	picking 116:7	255:6	238:4 324:7
perforation 292:15	permissive 247:15	218:18	platform 46:2	polyp 318:21 319:7
311:2	permitted 289:5	piece 178:13 182:3	platforms 45:18	319:18
<b>perform</b> 69:22	perplexing 49:10	216:1 269:8	<b>play</b> 203:20 215:15	polypectomy
87:13 91:7 186:3	55:14	286:16 300:19	242:5 344:22	318:14,17 320:9
198:11 222:8	person 6:5 97:2	<b>pieces</b> 286:9	played 123:14	320:10
232:20 277:2	193:11	piloting 9:9	playing 281:14	<b>polyps</b> 319:1,4,14
performance 2:11	person's 81:3	pipeline 9:4	<b>plays</b> 323:12	319:19,20
2:13,13,15 21:14	personally 53:18	<b>pitch</b> 104:3	pleasantly 144:18	<b>poor</b> 184:12 199:18
22:7,15,19 48:18	97:13 170:16	<b>place</b> 9:3 35:2	<b>please</b> 45:9 56:6	279:1
52:13 62:21 66:3	257:10 283:10	59:17 121:13	99:6 128:17	<b>poorest</b> 182:8
66:5,8 75:12	315:7	122:6 177:22	134:19 137:9	popped 44:9
77:12 78:2 79:21	perspective 81:3,15	182:6,14,15,16	141:15 212:1,3	<b>pops</b> 261:10
84:20,21 88:19	87:22 97:17	198:15 204:13	216:16 221:9	population 16:4
92:17 105:1,4,14	105:20 242:21	259:11 279:7	241:8 265:8 278:7	60:18 61:3 62:7
106:2,5,9 135:1	296:11	307:2	285:13 289:12	119:15 155:21,22
135:22 137:5,7,11	perspectives 84:17	placement 28:18	295:6 300:5 301:4	156:4 163:12,16
142:2 149:21	Petersburg 17:21	places 281:7,12	302:1 330:20	166:22 167:11,22
161:20,21 162:1,3	<b>PharmD</b> 1:13	plaguing 49:7	352:6	168:2,3,22 169:6
194:17,20 196:14	<b>phase</b> 111:18	<b>plan</b> 88:7 89:1,12	<b>plus</b> 324:4	175:6,12 177:3,6
196:16,20 214:13	<b>PhD</b> 1:11,12,14,20	94:18 112:8 115:4	pneumonia 40:22	181:7 184:19
214:15,16,18,19	2:2,14,20 3:2,3,7	126:5 242:19	191:13	185:1 186:18,19
224:20,22 225:8	phenomenon 311:4	250:8,10 334:16	point 15:11 22:2	187:17,19 188:8
225:12,15 227:18	<b>phone</b> 18:6 52:3	334:18 335:9	27:5 54:21 55:2,3	188:16 219:8
229:9 247:8 250:3	57:8 91:18 99:8	336:13,14 337:6	55:11 71:21	239:17,19 256:2
259:18 271:8,18	99:15 130:2 191:5	<b>plane</b> 12:16	109:14 123:9	256:10,12 277:6
271:22 272:4	220:4,6 265:22	planet 54:13	130:22 136:15,20	295:16 298:21
276:8 296:22	266:2,3,9 278:6	planned 29:8 31:2	160:14 172:10	299:19 301:18
297:3,14,19 327:7	278:16 332:14	31:8,20 32:6	175:1 204:21	312:13 316:18
performed 30:5	347:10 352:2	33:18 34:10 35:3	221:5 258:5	324:3 336:14
59:12 76:5 116:21	phrased 27:21	35:7 43:8 50:22	261:17 262:19	340:22
117:21 118:1,19	<b>physician</b> 290:13	74:14 140:1,5	269:10 277:1	<b>populations</b> 119:16
119:6 133:2	308:4 309:11	146:18 171:5	279:14 289:19	155:14,18 158:20
140:19 205:8	312:18 313:1,5	203:5 205:16	296:4 312:4 314:4	194:22 219:2
232:18 238:7	314:20 315:10,12	222:18 236:3,5,7	314:13 315:11	276:14 340:16
290:12,20 291:13	315:21 316:1	236:13 267:8	318:6 324:5	<b>portfolio</b> 96:19
<b>performers</b> 87:8,8	322:18	268:18 269:7	329:13 337:15	185:2 255:3
87:9 111:13,14	physician's 322:11	277:12 278:4	344:9	<b>portion</b> 39:10
performing 241:2	physician-level	305:7 313:16	pointed 125:8	160:21 187:4
322:8,18	306:14	332:22 333:3,7,15	pointing 168:21	portions 68:11
performs 185:15	physician-owned	333:16,17,21	PointRight 14:4	portraying 52:22
192:12 204:19	307:17	334:9 335:10	<b>points</b> 17:1 33:15	position 80:15
period 21:18	<b>physicians</b> 100:6	<b>planning</b> 194:4	83:1 283:3	187:11
156:15 203:1	307:18,20 308:9	335:5	<b>policies</b> 11:4,5,7	positive 254:17
243:19 294:3	<b>pick</b> 40:11 305:15	plans 40:12 112:2	<b>policy</b> 80:19 100:20	positively 251:11
350:13 351:13	<b>picked</b> 215:6 216:4	158:4 161:2	109:15 182:1	possibility 50:4

84:1 248:4	predicted 143:7	292:13,18	55:16,20 65:13	74:8,15,17,19
possible 24:15	149:1 248:18,21	prevented 317:12	146:1 152:7	75:2,10 99:5
50:18 52:13	282:20	Preventive 298:1	158:16,17 162:2	101:4,5 102:12,13
102:18 172:16	prediction 235:6	previous 156:12	165:21 169:11	102:18 107:14
247:18 281:5,6	predictive 203:16	198:5 201:11	173:6 181:16	114:4,22 115:3,4
296:10 313:9	predictors 276:14	207:11,13	182:3 231:22	116:2,4 117:21,22
possibly 58:15	preface 78:18	previously 74:7	234:10,14 235:8	117:22 118:4,6,14
190:17 318:13	preferences 315:20	76:8 138:20 242:1	239:9 241:15	118:18 119:5
post 13:17,20	prefers 336:14	269:9	242:16 245:4	124:16 127:6
311:11 327:4	prejudice 325:10	primarily 183:10	248:9 270:6 272:9	128:22 146:17
345:15	prep 291:6 315:14	188:11 221:18	310:16 316:10	156:13 233:5
post-discharge	315:15	256:4	325:8	248:10 290:17
73:9 336:11	preparation 165:11	primary 62:2	problem 35:12 49:7	291:15 292:2,5
post-meeting	174:17	140:16 141:18	51:14 52:7 55:10	299:20 309:21,22
349:11	preparing 180:22	166:3 180:4	62:7 70:20 124:17	332:6 348:21
post-op 307:13	present 1:11 2:17	181:17 182:10	153:8 182:4	process 9:14,20,22
potential 16:7	3:9 35:22 137:21	200:9 232:20	210:11 237:5	10:11,12,19,20
22:10 26:12,15	210:10 223:21	313:5,15,19 314:5	283:18 317:11,11	11:19,20 19:4
27:22 28:4 34:7	293:8 330:21	314:7 317:6	339:19	34:22 39:5 58:11
35:12 45:4 47:14	presented 43:18	principal 130:2	problematic 55:15	59:18 63:11 64:14
92:9 123:8 170:6	316:3 331:5	218:7 244:8	55:20 146:3	65:19 76:13 79:15
170:9 176:10	preside 351:16	principles 101:8	180:11,16	95:17 100:15
224:4 306:15	<b>President</b> 2:10,11	102:4	problems 49:11,15	123:17 144:9
322:2,15 341:17	presiding 1:10	<b>prior</b> 69:9 109:18	142:16 336:20	213:17 215:13
346:16	press 128:17 212:4	218:9 229:9,13	procedural 268:19	254:16 286:15
potentially 92:18	216:16 352:6	236:9 242:16	procedural-based	294:14 351:19
157:10 173:3	pressure 318:4	246:16 340:13	69:5	processes 127:17
177:19 217:11	322:18	346:12	procedure 4:10	174:17,21 194:4
235:6 253:1	Presumably 88:16	prioritize 192:4	15:6 29:5,8 47:14	202:19
269:20 292:13,17	pretty 53:8 62:6	prioritized 131:22	80:1 101:9,11	produce 78:1
346:11	83:20 93:10	<b>priority</b> 23:1,2,4,11	113:14,17 114:5	professional 45:11
<b>power</b> 124:14	118:21 149:10,11	23:12,20,22 24:3	114:12,16,18	proficiency 307:4
180:3 277:8	149:17,20 150:2	66:11,13,15,19,22	116:21 117:9,10	319:14
powerful 237:11	235:8 312:8,12,17	93:19 106:11,18	117:14 118:15	profound 338:17
276:14	341:20	106:20,22 107:4	119:18 120:10,16	350:18
practicality 336:1	prevalence 135:3	137:14 138:3,6,9	120:19 121:13	program 44:10
practice 37:21	prevalent 191:19	138:13 197:1,13	127:5 140:5 203:5	48:2 51:7,9 94:14
115:18 174:8	prevent 50:10 74:3	197:17,21 225:18	205:16 230:8	94:16 106:17
314:7 322:5	170:15	225:19 226:10,14	233:3 291:1,7,13	112:22 113:1,2,4
practices 171:15	preventability	272:7,9,15,19	291:21 298:8	113:8 130:14
pre-surgical	132:9 172:11	297:21 298:13,16	303:5,11 304:3,16	143:2 160:5,7,11
290:13	173:4 207:17	private 7:4	305:18 310:12	160:17 214:2
pre-work 87:7	217:15,20 218:3	privately 168:11	311:11 312:1,19	267:1,3 308:22
precise 126:14	preventable 100:3	privilege 351:17	313:12 325:4	programs 7:3,4
precision 201:20	170:11 171:4	probably 7:17	333:2	10:3 36:6,17
<b>predict</b> 231:17,20	172:13 173:3,13	23:10 29:9 37:6	procedures 44:11	82:20 84:8 94:11
301:17 302:4	217:11 262:14	41:18 49:8 53:6	51:16 59:5 60:8	98:5 113:9 159:8

	1	1		
165:9,13	100:6 236:12	pull 80:14 115:17	84:7 85:5,7,10	140:22 141:13,17
progress 327:19	291:17 292:20	149:4 304:20	91:8 92:21,22	142:20 143:14,20
progressed 223:11	293:3 306:17	305:6	93:3 100:7 113:1	146:10 147:6,12
project 2:12,13,15	317:20 318:22	pulled 305:3	113:2 121:17	151:6,20 152:19
274:22	319:9 322:6	pulling 279:5	122:7 124:7	153:15 155:12,15
projects 266:20	327:10	<b>pump</b> 30:16	125:18 126:16	155:21 161:17
prominent 36:6	provides 294:7	pun 289:2	127:16 130:13	164:6,20 167:2
promotes 254:16	providing 125:21	punctures 101:12	133:7 134:2	169:1 173:17
<b>proof</b> 249:21	178:5 257:22	<b>purely</b> 97:17	144:13,20 145:20	178:17 183:2,9,12
proper 34:13,14	312:22	328:17 339:18	146:6 154:3 156:2	183:19 186:9
50:19 51:5 78:9	<b>proxy</b> 210:13	purpose 10:9,10	158:14,15 178:8	187:21 203:18
properly 224:6	341:14,15	11:21 215:5	183:17 193:1,8	206:15 210:3
property 261:15	psychiatric 2:5	purposes 88:17	195:2 199:16,18	213:3 239:20
prophecy 69:12	336:17,20,22	89:22 91:9 133:18	202:18 215:1,9	247:12 249:17
299:8	337:3,4,5,9,14	213:15 215:8	243:17,20 255:17	256:17 257:2
prophylaxe 248:8	<b>public</b> 4:15 5:6	245:22 335:22	256:3,6,7,14	262:2 276:3,10
proportion 25:13	14:3 45:7,7,15,20	<b>purview</b> 200:3	261:6 265:12	277:14 278:12
73:14 91:5 173:2	46:12,18 87:16	212:17	266:22 291:18	281:5,16 303:13
187:7 243:19	92:18 93:4 127:13	<b>push</b> 83:6 254:13	292:7,9 293:11	310:21 312:5
327:9 333:14	140:15 176:4,15	261:16 289:12	298:9 303:3,8	313:2,8 316:7,22
341:10 342:7,22	203:6 211:16,16	<b>put</b> 30:18 42:14	307:3 311:15	318:12,13,16
proportions 327:8	211:20 212:3,4,6	53:10 89:10	313:14 316:1	323:15 347:5
proposed 303:8	215:3,10 216:14	109:15 122:4	319:2,5 339:3	questioned 37:20
proprietary 45:4	216:16 220:6,8	177:22 203:8	quantify 77:17	175:21 247:3
46:13 64:3	223:7 242:16	244:2 245:7	84:3	<b>questions</b> 11:11,13
<b>pros</b> 59:16 252:8	244:1 245:18	259:11 273:16	quantitative	12:11,19 20:16
prospective 308:20	246:21 248:22	283:6 307:7 318:7	253:22	21:7 32:12 46:20
<b>prove</b> 269:13	251:4 255:7	322:17 335:6	Quantitatively	50:2 56:1,16
<b>provide</b> 37:16 54:2	256:17,19 257:3	337:20	78:4	71:16 95:1 97:16
70:6 85:1 91:19	257:13 259:14,22	putting 161:19	quarterly 125:22	102:21 110:15
151:18 154:9,11	266:21 280:22	254:17 339:15	126:6,7 208:7,15	128:1 145:8
165:9 179:8 188:3	282:8 283:12		208:19 259:10	150:17 179:13
243:11 293:7	294:3 322:9 335:7	$\frac{\mathbf{Q}}{\mathbf{Q}\mathbf{R}\mathbf{A}}$	quarters 292:3	198:13 203:14
294:2 327:17,19	341:9 342:16	<b>Q&amp;A</b> 91:14	query 53:9	221:21 223:14
329:4,8	344:7 349:18,20	qualified 294:1	<b>question</b> 29:15,20	270:2,9 285:5
<b>provided</b> 24:9 36:3	351:13 352:2,3,6	<b>qualifier</b> 263:7	32:14,16 34:1,6	300:13 303:2
36:9 63:3 121:15	<b>publicly</b> 11:6 45:17	345:16	38:12 40:17 41:10	332:12,15,16,19
123:2 126:8	45:22 80:7 90:12	qualifiers 343:9	43:6 45:3 63:9	337:17 340:11
142:21 160:5,6	91:11 130:9 131:6	<b>qualify</b> 234:20	64:2 65:11,20	343:20 347:7
165:9 178:20	133:21 242:8	333:3	71:22 72:15 73:5	quick 6:11 11:10
196:4 256:7 299:3	245:9 248:19	qualitative 253:22 qualitatively 78:3	78:15 82:7,22	13:8,15 41:9 48:8
300:20 327:21	255:2 256:22	quality 1:1,8 3:5	83:12,18 84:6	150:16 155:12
provider 114:6	258:7 259:13		87:6 91:17 97:7	165:18 193:13
293:9 313:16	260:5 268:6	4:12,14 7:10 8:10 28:21 53:9 74:2	97:13 113:10	220:10 270:7
315:10 318:5	321:17 331:19		115:8 117:17,19	276:22 281:4
320:1 325:11	<b>published</b> 39:6	75:11 79:7,7,7 82:4,4,18 83:20	118:13 122:10	290:6 316:7,15
providers 55:7	171:1 252:11	02.4,4,10 05:20	123:7,7 124:3	327:19 348:8

<b>quicker</b> 278:14	randomization	348:20 349:4,8	reach 13:22 316:8	199:16 200:12
quickly 13:9 15:3	171:17	rated 88:2	reached 14:10,13	201:1,12 205:13
17:10 118:9	randomly 68:10	rates 4:16 29:14	14:16 15:2 60:5	205:15 211:12
123:18 124:6	158:7	80:1 105:6 127:20	303:19	218:10,18 219:15
125:13 276:16	range 21:21 22:3	130:18,18,21	read 300:14	222:3,7,14 223:1
278:8,12,22 282:2	52:13 63:4 68:18	131:1,2 133:6,15	reading 322:21	223:6 225:2,3
293:13 331:14	70:12 105:10,11	136:19 154:2	readmission 4:5,7	236:3,5 244:9
338:8	105:21 106:1	155:1 156:7	4:9,11,13,16,19	248:6 252:12
<b>quite</b> 6:7 39:6	131:16 135:6	159:12 160:8	13:17,20 14:7,8	253:18 254:1
124:6 136:16	136:19,22 149:16	166:17 170:21	14:18 18:19,21	264:20 267:2,8
142:14 147:19	154:10 195:15,17	171:16 173:22	21:17 23:7,16	268:3,14 269:7,18
155:22 158:21	196:10 199:9	174:22 177:16,18	29:12,13 31:10	272:10 277:12,17
165:7 199:10,15	200:4 203:17	177:20 179:9,10	36:2 38:1,4,22	278:4 288:8
201:10 217:1	204:12 219:16,17	186:8,13,17	40:20,21 47:20	299:14 302:14
297:5 312:14	297:4 309:22	187:11 192:18	57:3 58:22 61:5	304:11 306:5
<b>quorum</b> 289:6	310:21 311:1	208:19 218:10,10	62:7 63:4 68:13	332:22 333:4,6,19
quote 228:11	ranged 198:7	219:15 222:3,7	69:1,2 70:2 73:12	333:22 334:2,9,10
	<b>ranges</b> 135:7	223:2,6,8 225:3	73:18 74:4,20,21	334:16,17 335:11
<u> </u>	ranging 295:21	244:15 248:6	75:2 76:15 77:10	338:16 339:13
<b>R</b> 282:1	Raphael 2:2 8:5	252:12,17 253:10	77:14 78:8 80:1	340:2,5,6,7,9
<b>Race</b> 37:15	rapid 8:22 48:2	253:19 254:1,9	90:13 96:6,21	346:2,3,8 348:11
race/ethnicities	259:7	264:20 269:18	99:4,22 100:4	348:14,16,18,20
138:1	rapidly 253:14	272:11 282:21	105:6 108:19	349:1,4
race/ethnicity	277:19	295:19 299:12	113:13,17 114:17	readmissions 1:3
179:16	rare 31:4,5 51:15	303:10 305:14	119:13 120:12	5:2 6:4 37:2,4
radiologists 100:18	51:16 157:18	320:9,12 336:11	121:5 125:3 127:3	59:4 66:15 100:3
<b>raise</b> 289:6 306:22	rate 4:5,7,9,19,21	338:9,16,18 349:1	127:12,20 128:21	106:17 107:10
raised 44:14 48:5	18:19 21:17,20	ratio 14:9 36:2	130:7,12 131:1,20	126:22 130:6,16
49:20 66:3 86:2	29:12 31:10 57:3	143:11 149:14	132:2 133:3,6,14	131:5,9 132:7,10
86:17 98:12	65:7,8 68:13 73:8	rationale 21:5	133:22 134:8	137:16 139:22
156:17 183:10	73:18 74:20,21	28:19 30:1 171:9	135:3,15,17	143:7,13 149:2
185:18 235:10	75:2 76:16,22	173:21 174:13	136:19 138:1	150:1 151:21
330:16	77:6,10 89:7 96:6	195:8 200:2,18	143:4,6,9 146:2	156:13 172:11,13
raises 65:20 282:1	99:4 125:21	270:20 290:7	148:7 150:22	172:14,20 173:2
Ranasinghe 3:3	127:12 128:21	294:7 309:20	154:2 155:1 156:7	173:14,20 174:9
289:16,17 290:4	135:4,8,15,17	rationalization	159:12 161:6	175:11 178:4
294:20 296:1	143:6,10 148:7,19	118:21	166:17 167:6	181:22 191:10,20
301:5,6 303:12	181:20 191:21	raw 126:1,3,7	168:11,20 170:10	192:1,12 193:22
309:17 310:20	192:16 195:11,18	re-approval 123:22	170:11,15,21	194:2,3 197:7
313:7 318:15	196:6,9 201:1,12	re-approving	171:16 173:22	202:20 203:13
319:16 323:14	236:8 244:3	249:20	174:3,22 175:13	213:9 217:4,10,15
<b>RAND's</b> 69:11	271:16 277:18	re-disclosed 39:12	176:3 177:19	218:3,5,16 222:12
<b>random</b> 25:15	288:8,19 291:14	re-endorsement	178:8 179:9,10,17	222:16,18 226:2,4
67:18 135:12	299:20 303:6	266:18	182:20 183:17	236:13 248:10
230:1,6,7 231:12	319:8 326:3	re-endorsing 124:2	186:8,13,17 190:9	252:5 262:7,13
231:13,21 232:2,3	333:19 334:2	re-introduce	191:21 192:5	266:13 267:6,9
232:5 273:3	339:13 348:11,14	330:20	195:11 196:6,9	268:17,19 270:3
				,

270:17 282:17	97:14 98:8,11	295:12 304:10,19	205:7 220:18,19	74:6 101:1 102:6
284:3 327:1	111:13 114:1	307:9 308:14	330:21 352:14	310:10
333:14,16 335:12	116:16 118:16	reassess 183:5	recorder 265:9	register 203:7
readmitted 36:13	119:11,15 121:12	reassurance 53:19	recording 17:4	332:13
70:19 91:3,6	122:7,11 123:18	reassured 270:2	records 144:19	registries 51:11
118:1,8 121:19	126:2 136:11	reassuring 77:22	<b>recuse</b> 103:5	258:15
151:1 152:4 175:8	142:13 146:10	recalculated 299:7	redid 339:11	registry 39:19
183:16 194:9	147:5,15 154:7	recall 109:7	<b>reduce</b> 48:10 65:16	59:19 60:11 63:22
240:20 244:8	158:11 160:11	recalling 341:6	65:21 131:5	64:16 65:19 79:18
332:5	169:1 178:10	recap 4:2 13:15	170:21 177:18	80:13 82:16 83:3
readmitted/not	207:11 208:15	receive 83:5 90:15	254:9	83:22 84:4 94:13
70:19	212:16,19 213:7	244:4,6	reduced 174:9	222:22 235:15,20
ready 21:7 22:12	213:12 214:2	received 68:16	reduces 277:17	240:22 241:3,14
23:20 33:4 43:3	216:6,7,8 218:8	203:6 223:5	reducing 9:13	241:20 245:17
43:19 46:21 56:1	218:21 232:16	245:15,19,21	130:5 236:15	255:8
56:17 104:16	236:18 237:17	294:14 331:15,21	253:18	registry-based 59:3
106:2 108:1	238:16 242:7,9	331:22 332:2,11	reduction 267:3	76:17
110:15 128:2,11	245:7 248:7	receives 90:11	reductions 202:19	regression 231:16
128:13 134:15	250:20,22 251:18	receiving 200:1	redundant 63:6	261:10
137:4 138:8	252:6 256:8 263:6	336:8	234:11 239:11	regularly 307:19
139:10 156:22	263:19 269:15	reciprocal 82:14	reengineering	rehab 13:18
169:19 186:6	282:2 283:8 285:8	recognize 7:6 92:4	10:15	rehabilitation
189:12 196:13	289:22 291:8	93:9 143:16	reevaluates 214:6	151:19
197:14 206:22	292:1,16 293:10	263:16	reference 148:21	rehospitalization
208:22 209:7	303:14,14,16	recognized 59:15	160:9 231:11	14:20
211:3,15 215:2	305:18 309:9	recognizing 8:13	referenced 69:11	rehospitalizations
271:1,20 272:13	310:1 317:17	162:2 188:18	258:9	14:5,14
273:8 274:5	318:3 322:6 336:3	208:18	referring 88:13	reintroduce 190:19
279:18 280:10	338:17 342:2	recommend 43:16	reflect 28:2 313:13	relate 39:2 94:7
286:21 287:22	343:17	238:1	317:4,5	related 51:12 94:10
307:14	realm 67:4 150:20	recommendation	reflected 258:11	96:13 98:14
<b>real</b> 123:7,13	<b>realtime</b> 216:11	16:1 69:8 74:8	reflecting 204:11	121:13 156:7
124:17 145:1	<b>reason</b> 29:11	75:9,16 213:5	reflection 53:7	181:16 202:19
158:2 183:2	141:18 142:1	298:1	reflections 16:13	221:21 291:6
217:20 235:13	144:11 229:22	recommendations	27:16 351:11	304:14 305:18
253:7 323:12	230:18,22 259:6	75:14 96:12	refreshing 277:21	308:1 311:22
<b>realize</b> 10:16	304:22 307:7	210:20 211:1	regard 65:19 133:3	327:20 328:20
212:10 217:11	310:1 313:20	347:3	192:13 208:20	345:6,18,18 346:9
234:17	317:19	recommended	254:5 334:13	346:10
realized 187:22	reasonable 25:6	214:7 284:18	337:16	relates 94:8 197:5
really 6:18,20 7:11	33:1 44:8 80:17	recommending	regarding 50:20	345:13
7:18,21 8:11,20	140:9 203:17	298:2	61:6 99:20 141:13	relating 339:20
8:22 9:17 11:5	262:11 271:15	reconsider 123:22	141:17 142:20	relationship 61:20
24:15 30:18 40:19	275:6,7	reconstruct 228:21	147:12 160:14	67:21 179:15
54:4,22 61:21	reasonably 341:19	reconvene 220:16	173:21 198:19	315:10 337:21
80:2 83:9 84:5	reasons 48:12	record 39:22 48:21	<b>regardless</b> 245:17	relationships
87:11,15 93:15	191:16 291:9	129:9,10 190:20	regards 69:11 73:9	188:14,18

	1	1	1	
relative 210:5	remember 185:3	47:17 50:4 79:4	265:19 281:15,18	84:6
relatively 34:21	233:8 276:3,10	88:17 89:21 91:1	researcher 129:18	restrictive 247:14
51:15,16 154:7,11	329:17	113:1,2 126:19	researchers 144:6	result 68:5 147:17
223:12 244:22	remind 35:18 61:4	127:13 140:15	resection 305:8	149:3 192:7,20
273:5 292:5	96:10 112:6	215:3,9 221:20	residence 152:2	236:7 251:15
release 92:18 93:3	156:11 265:8	223:8 241:18	resist 97:5	277:7 284:7 327:9
176:11	reminding 339:22	242:16 243:19	resolve 55:17	resulted 319:8
released 172:5	remodeling 235:16	244:2 245:18	resolved 181:8	resulting 133:8
releasing 93:4	removal 341:2	246:16,21 249:1	resource 259:2	results 5:1 15:21
<b>relevant</b> 147:21	remove 75:10	251:4 253:8 255:7	respect 17:2 20:19	26:2,6,11 46:8
336:14	319:1	256:17,19,21	25:12 27:8,9 28:9	94:12 150:3
reliability 23:1	removed 333:4	257:3,8,13 258:22	37:2 52:18 65:18	202:11 207:6
24:7,10,19 25:13	339:7 340:21	259:18 260:5	104:4 180:12	208:14 209:5
26:3,13,16 27:8	removes 319:7	266:21,22 268:4	239:7	211:12 250:12
27:17 28:1 33:6,9	removing 318:21	280:22 282:8	respected 45:20	288:9 298:16
35:22 52:21 53:1	319:3,4,18,20	283:12 331:20	51:21	299:16 300:7
53:5,13 67:4,6	renal 42:1,4	reports 46:1,4	respectively 26:9	320:18 321:10
70:8 71:6,8,11	<b>repeat</b> 17:14	223:5 244:1,6	79:12 105:7 203:3	324:20 326:4
85:18,20 107:7,17	rephrase 285:14	245:19 258:1	respiratory 4:13	327:5 331:19
108:1,4,8 133:9	replacement 84:21	304:4	15:7 191:9,15	332:9
133:15 138:17,21	report 45:18,19,22	represent 60:12	192:2 193:20	retest 68:3
139:1,7,10,12,16	46:2,10,17 52:9	75:4 76:15 158:6	194:9 195:12	returning 327:22
154:22 187:10	52:12 76:2 77:14	representatives	196:2,11 197:11	reveal 185:11
192:14,21 198:3,7	85:1 90:12,15	221:7	201:16 204:5,14	revealing 89:20
198:14 201:6,7,9	92:19 100:1 126:9	representing	204:15 208:1	reveals 63:14
201:13 202:2,4,6	127:11 208:14	221:11,16	211:11 219:7,14	review 5:1 103:10
202:11 226:17,19	243:14 244:12	represents 27:22	348:17	103:22 140:8
226:19 227:1,9,13	245:16,21 248:19	64:16 76:21,22	respond 38:10	205:10 222:2
272:22 274:7,11	255:2,11 258:3,10	79:18 224:2 226:3	40:16 125:10	260:13 269:5
298:19,20 299:22	258:16 259:13	241:15	141:11 165:2	274:22 318:4
300:3,7	267:21 275:20	reproducibility	208:10 350:5	326:8 328:16,21
reliable 80:9 85:21	323:2 330:11	24:20 27:6 32:11	responded 73:5	329:10 350:5
198:10 299:15	331:18 349:17,18	32:12,17,19 33:4	response 69:7	351:7
308:13	350:2	70:9	84:10 85:21,22	reviewed 63:1
<b>relied</b> 304:9	reported 26:6	reproducible 53:9	156:20 162:18	73:21 295:19
<b>rely</b> 103:20	45:17 77:13 80:7	request 43:9 329:9	261:12	347:21
<b>remain</b> 55:14	88:8 89:16 90:13	requested 327:3	<b>responses</b> 12:13,19	reviewer 103:11
113:12 238:12	90:21 91:11	require 35:5	responsibility	reviewers 17:2
remaining 49:9	126:20 242:8	required 193:4	250:2	reviewing 6:7
63:21 65:16 336:5	243:12,20 245:9	329:12	responsible 100:2	350:20
remains 53:15	256:22 258:8	requirement	155:5 346:22	<b>reviews</b> 205:8
183:22 231:3	259:9 267:17	198:21 241:18	responsive 80:6,22	226:21
271:10 272:9	268:7 269:16	requires 29:1	261:4	<b>rich</b> 195:7
<b>remarks</b> 247:22	273:4 304:6	research 3:7 39:13	responsiveness	<b>ride</b> 310:6
252:20 254:20	317:18 321:18	57:19 132:8	281:9	right 7:18 25:5
300:10 320:21	<b>reporting</b> 26:2 45:7	167:20 172:21	<b>rest</b> 256:4	27:13 32:16 34:2
321:13,14	45:16,20 46:13,18	219:10 239:15	restating 83:18	38:18 39:1 81:10

			1	
83:9 88:3,5 93:16	318:14 348:13	<b>run</b> 5:1 15:21	129:22 191:3	155:17 157:17
100:5 101:8	risk-standardized	164:10,15 169:12	schools 264:7	178:3 183:5
112:21 121:10	4:7,16,21 18:21	169:12 188:2,5	Schuster 3:4 130:3	185:14 193:22
127:2 146:21	21:20 68:12 76:15	275:22 327:4,5	191:5	194:8 208:8 214:3
154:3 160:21	77:6,9 99:4 105:6	331:12,16,17	Schwalenstocker	214:15,17,21
168:1 169:19	128:20 135:7	332:9,12,19 334:5	212:8,12	219:18 237:6
178:22 179:2	195:18 222:3	running 165:11	science 7:20,22 8:9	243:1 246:15
184:2 209:20	223:2,6 244:3	<b>runs</b> 230:10	scientific 9:16	249:15 251:1,4,6
214:5 232:13	264:19 288:7,18	rural 151:5	22:22 24:6 67:4	251:9 252:16
242:3,6 251:21	326:2 339:12		107:6 138:16	264:9 265:9 273:4
259:9 275:19	348:20 349:4	S	198:2 221:20	275:13 277:2,5,7
279:15 294:20,21	<b>RN</b> 1:14,16 2:1,14	safety 48:13 192:22	226:16 251:8	277:19,22 278:5
296:1 306:4	road 97:5 209:13	286:4 342:10	272:22 298:19	291:14,15 305:9
318:12 349:15	<b>Robert</b> 3:2 103:2	<b>sake</b> 97:14	scope 83:22	308:6 309:6,8,10
rigorous 24:14	<b>Roberts</b> 2:2 115:22	same-day 120:9	score 299:16	311:11 312:6
rigorously 34:17	345:4	sample 25:18 52:7	323:18 331:22	325:1 329:22
<b>rises</b> 70:1,1	robust 68:6 262:21	61:13 67:7 68:3,7	score-out 25:7	330:1 335:1 338:4
risk 4:9,18 20:7	264:4 287:17	69:4,17,19 80:12	scores 123:3 133:9	340:8 342:18
36:1 47:20 49:20	341:7,12,16,19	133:8 148:11,12	318:3 331:22	343:5,6 352:2
50:4,5,8,10,13,17	robustitude 287:20	187:2 192:15	scoring 261:14,22	seeing 37:18 62:14
50:19 51:2 58:17	Rochester 1:19	198:11,12 208:20	<b>screen</b> 278:4	66:2,17 70:9 71:5
59:20 73:11 75:18	<b>rock</b> 182:6	273:3,15,21	screening 292:3	81:18 85:2 86:2
75:20 76:4 79:16	<b>role</b> 330:7 345:1	299:15 300:15	298:3	95:21 145:18
92:10,11 96:5	roll 115:2 190:1,1	301:9,11,12 323:9	scroll 278:1	201:22 202:3
102:6 116:1,7,10	rolled 131:13	323:16,17,19	Sean 3:2 17:22 18:1	226:8 227:7 235:9
116:12,19 119:13	158:12 219:5	343:3	25:10 27:4,9,21	240:7 247:9 249:3
121:12,21 140:10	<b>rolls</b> 177:11	samples 67:14,18	39:15,20 40:16	269:17 274:2
161:5 168:11,20	<b>Ron</b> 293:15	68:9,14 69:18,21	52:1,5	299:18,22 303:10
179:16 187:19	<b>RONALD</b> 2:6	SAS 160:5	season 208:15	320:13 321:4
188:20 204:13	room 1:9 76:20	saved 351:5	seasonal 208:1,7	324:14
217:4 218:6	84:15 92:10 99:8	saw 26:20 105:20	second 12:4 36:1	seen 7:17 105:14
219:19 224:22	193:11 212:7	167:22 199:10	100:9 141:10	106:1 138:22
225:4 228:8	220:8 289:5 315:4	218:21	155:16 177:15	181:20 238:8,9
231:17 235:6	340:13 341:6	saying 78:18	185:6 229:3	246:22 248:20
248:18 253:9	344:21 352:4,9	212:18 238:7	260:10 284:10	251:14 281:2
263:22 290:16	<b>Ross</b> 1:17 62:2 67:5	284:2 293:13	<b>section</b> 112:4	310:7
301:1,3,14,15,20	71:14 86:12 104:8	314:18 317:9	sector 7:4	Select 1:20
318:18 322:2,13	295:2 314:14	says 35:6 81:4,7,14	Security 38:19 39:2	selected 158:7
322:16 335:17	316:5 318:11	SCFES 2:2	39:4,7,9,11 245:2	316:18
336:12 338:15,21	321:13 325:2	scheduled 268:18	see 11:10 17:5 25:2	selecting 47:19
348:10,22 349:7	346:6	313:17 344:14	42:15 44:18 47:10	345:19 346:11,15
risk-adjust 8:2	rough 228:10	349:12	48:2 53:6,16	346:18 347:5
20:12	roughly 23:6 70:12	scheduling 351:6	69:14 70:13 71:1	selection 302:17
risk-adjusted 4:4	81:9,11 226:3	scheme 235:7	76:10,20 77:5	selects 10:1
14:22 15:5 18:18	257:9 324:12	schizophrenia 337:1	83:7 84:15 86:17	send 345:7 346:8
57:2 125:21	<b>RSRR</b> 4:7,9,19	<b>School</b> 1:21 3:7	92:8 95:2 113:8	346:12
127:18 294:9	<b>rules</b> 12:5,8 120:18	SCHOOL 1.21 5.7	125:12,22 150:20	sends 331:17

senior 1:17 2:11,12	151:2 152:6,10,13	21:9 22:14 23:21	<b>shift</b> 261:21	72:13,21 78:3
2:14 221:19	152:14 159:17,18	33:5 43:20 46:22	shifts 122:6	87:5 91:1 122:2
sense 109:5 114:2	160:9,9 162:6,8	56:3,19 62:16	shining 325:5	146:12 147:15
115:19 134:13	162:12 163:13,20	66:4,18 71:7 86:4	<b>SHIPPY</b> 2:4 41:9	149:6,11,20 150:3
135:10 141:22	166:3,5,11 178:22	86:18 95:3,22	41:17 347:9	150:9 167:12
142:4 154:13	179:19,20,21	104:17 106:4,21	shock 229:21	171:7 190:17
160:12 173:13,16	180:1,3,4 186:22	108:3 110:3,17	<b>shoot</b> 263:14	191:7 192:9
174:14 179:8	188:4,6,11,13,17	128:3,14 134:17	<b>Shore</b> 1:18	199:15 203:11
189:4 200:22	199:14 202:22	137:6 138:9	short 128:1 223:13	275:1 302:14
204:17 288:20	213:11 217:1	139:11 157:1	<b>show</b> 40:13 237:14	307:2
sensitivity 79:9	227:20 250:9	169:20 189:14	239:3 281:14,18	similarly 59:5
80:2 203:1 260:22	277:5 299:10	190:4 194:13	showed 109:1	133:2
261:5 281:7 341:3	345:21 346:4	196:15 197:16	167:20	simple 34:22
sent 118:9 332:9	sets 20:2 145:1	202:5 206:3 207:1	showing 256:1	simplicity 336:1
338:2,6	159:21 179:12	209:1 211:6	shown 177:7	simplistically 19:19
sentences 50:1	276:12 277:3	224:15 225:11	202:21 239:16	simply 35:5 153:10
sentiment 284:12	344:19	226:9 227:8 240:8	281:20	257:5 292:22
284:21	setting 113:16,18	243:2 249:4,10	shows 177:7 295:19	sincere 15:13
separate 28:22	114:12 119:18	264:11,17 271:2	Shriner's 151:14	single 34:20 142:6
30:13 68:7 98:18	231:15 291:13	271:21 272:14	shrinkage 55:5	148:7 230:13
122:19 255:12	292:20 302:12	274:6 279:20	147:18 149:19	238:16
September 350:11	304:18 316:9	280:11 286:22	sick 303:11 304:13	sit 6:16
serious 292:12	317:14	288:2 296:17	side 77:21 202:1	site 34:19 322:8
317:11	settings 112:19	297:13,18 298:12	262:10 282:16	sites 34:19 40:2,6
serve 18:3 130:14	122:8 184:21	300:2 320:14	310:12,21 329:7,7	63:16,21
131:20 134:1	217:22 290:14	321:5 324:15	345:8 346:8	sits 302:7
219:2	307:21	325:16,22	side-by 345:7	sitting 17:6
service 37:12 91:15	seven 100:2 106:15	<b>Shahian</b> 3:3 18:7	sideways 42:15	situation 30:21
91:22 115:13	138:13 297:19	57:6	<b>Sigma</b> 10:15	80:10,16 330:3
148:5 168:5	310:11,11,13,16	shaping 293:21	signal 25:14 26:4	situations 51:20
312:22	310:19 311:7,13	<b>share</b> 70:4 77:18	48:13 311:16	314:5
services 129:18	313:6 316:19	179:4 219:21	313:14 319:2	six 9:6 10:15
181:19 182:11	seven-day 288:18	233:9 258:4	signal-to-noise	137:16 243:7
298:1	311:14	273:17 286:1	24:15 32:21 53:1	273:14
SES 210:9 213:6	<b>seventeen</b> 24:3 44:3	341:8 342:14	53:14	<b>sixteen</b> 107:4
215:13 327:8	106:9 110:21	sharing 330:8	significant 47:18	227:13 274:11
341:22	139:16 225:16	<b>sharpen</b> 190:16	162:17 171:8	<b>sizable</b> 187:4
set 21:18 25:20	Seventy-two	<b>shed</b> 52:20 83:7	195:19 225:7	size 61:13 69:4
27:13 28:3 31:14	332:10	shelf 81:18	234:20 235:13	153:9 198:11,12
37:19 61:3 68:15	<b>severe</b> 35:10	<b>Sherrie</b> 1:9,12 6:14	239:16 271:18	208:20 273:15,22
82:16 83:1,2	204:14 218:14	11:10 103:8 237:1	275:12 276:7	299:15 323:9
87:11 93:5 108:17	severity 269:21	260:20 305:21	291:16	sizeable 160:20
115:14 124:15	sex 107:13	328:7 341:21	significantly 92:19	173:2
127:7 132:12	shades 32:8	351:8,15	98:12 171:20	sizes 52:8 80:12
136:16 143:22	shading 161:14	Sherrie's 83:18	<b>signs</b> 292:6	133:8 142:17
144:6,12,17	169:15	84:6 312:4,10	similar 10:5 62:12	148:11 187:3
148:21 150:4	SHAHAB 2:15	<b>shied</b> 238:6	66:12 69:10 71:17	192:15
	I	I	I	1

abill 206.11 207.4	gagiadamagraphiag	216.15 202.10	53:20	standardization
<b>skill</b> 306:11 307:4 <b>skilled</b> 14:6	sociodemographics 37:15	316:15 323:12 328:3 330:2 336:1	specifications 38:7	148:17 149:9,12
skills 43:12	socioeconomic	336:19 337:6	90:9 164:15 165:8	148.17 149.9,12
<b>Slattery</b> 3:5 221:17	145:16 177:4,13	sound 234:15	185:21 234:3	standardized 4:9
221:19 241:4,9,10	180:13 210:11,14	sounded 181:7	specificity 78:5	4:19 14:8 36:2
245:14,14 254:19	337:17 338:10,22	275:1	203:2	96:6 253:10 297:4
257:15 260:3	339:2,10 341:17	<b>sounds</b> 274:20	specified 24:17	338:16 348:11
slight 275:9	solid 51:7	320:8	36:10 37:13 43:8	349:1,8
slightly 59:2 78:13	solution 50:5 51:3	<b>source</b> 60:13,20	43:15 53:14 167:9	standardizing
118:8 274:19	<b>solution</b> 30:3 31:3 <b>solutions</b> 35:12	61:1 82:5 97:21	185:5,9,20 186:1	148:19
276:11 277:18	somebody 185:14	98:9 205:20	229:6 238:14	standing 9:5
334:1 338:15	261:15 278:6,11	<b>sources</b> 81:8,11	specifies 127:2	212:10
339:12	319:6	147:3 176:7	<b>specter</b> 282:1	standpoint 62:11
slowing 306:1	someday 210:19	space 174:12	spectrum 74:2	87:12,17 88:12
slowly 281:18	someplace 314:6	spanning 122:7	speed and 74.2 spend 240:1	183:14 231:1
small 36:22 48:16	somewhat 19:6	spare 10:18	spending 15:16	233:3
52:7 61:12 73:14	20:8 53:12 137:2	speak 59:1 64:8	spent 253:14	<b>Stanford</b> 1:21
77:16 89:9 133:7	140:1 184:11	71:21 72:6 73:1	350:20,22	staring 81:10
154:8,12 175:9	232:11 234:11	82:7 88:14 105:1	<b>spirit</b> 6:20	180:20,21
192:15 235:8	soon 259:12 265:7	134:6 148:15	<b>split</b> 67:7,14,18	start 12:15 147:8
237:14 244:22	sooner 215:19	254:5 265:7 278:7	68:3,10 69:18	172:10 182:2,15
282:11 318:13	sorry 23:1 48:21	speaking 64:22	298:21 300:15	182:15,16 198:2
323:9 334:1 337:8	58:4 78:11 119:21	199:1 221:18	splitters 114:21	223:16,20 255:9
337:13 340:12	141:16 146:18	262:22	splitting 115:9	262:10 290:6
small-volume	171:5 189:8	speaks 138:6	148:12	295:4 325:5
61:10 147:22	221:15 233:8	Spearman-Brown	<b>spot</b> 210:21	326:13,13 349:17
149:4	241:9 248:21	69:12 299:7	spread 22:3	350:13
smaller 115:6	249:9 253:16	Special 2:10	spring 331:10	started 13:4,6
186:12,13 201:17	279:12 282:3	specialize 307:17	St 17:20 229:20	185:19 193:17
241:16,17 334:3	308:19 320:21	specialties 100:16	stability 68:5	starting 168:4
smallest 182:9	sort 10:8 27:18	specialty 150:18	stable 80:8 186:12	starts 104:19
343:3	48:10 81:2,16	151:7 332:1	258:8	189:17 211:8
<b>smiling</b> 95:14	82:14 109:3 118:6	<b>specific</b> 90:12	staff 2:9 10:14,14	225:14 227:10
329:18	122:6,17 124:4	131:19 151:9	10:19 11:1 18:9	240:10 243:4
<b>SMITH</b> 2:5 161:16	142:13 145:20	169:6 175:17	stage 10:10	249:7 264:13
162:15 163:2	151:12,16 153:7	209:10 238:8	stages 253:20	274:9 280:13
<b>SNF</b> 14:4 96:21,21	154:2 161:8 164:5	244:1,6 319:20	stakeholder 131:15	287:3 288:4
346:2,3	177:11 178:12	321:14,15	stakeholders 131:4	296:18 297:15
<b>Social</b> 38:19 39:2,4	186:20 190:14	specifically 60:12	<b>stakes</b> 248:9	300:4 320:16
39:7,9,11 245:2	201:21 213:4,17	98:4 118:12 130:8	standalone 36:18	321:7 324:18
<b>Society</b> 2:19 3:3 4:5	233:6 237:20	175:6 179:11,14	standard 79:19,21	325:18
18:6,9,19 19:8	244:12 246:17	218:11 248:16	81:4,6,6 107:12	state 8:8 17:3
20:4 34:16 45:10	248:20 250:1	262:8 277:4 302:3	107:16,18 108:21	129:14 141:14
67:12 76:5	283:16,20 285:4,5	310:19 311:20	135:13,16 183:5	145:3 151:1,22
sociodemographic	303:9,21 306:1	327:6	196:5,8 205:1,7	152:1,4,16 155:14
8:2 161:9 209:11	309:18 310:3	specification 28:9	205:11 238:4	156:6 157:13,13
209:15	311:12,15 313:3	28:16 29:17 41:10	312:14 342:11	159:5,5,12 160:6

Г

160:11,12 162:8 162:10 164:8,12steering 1:3,8 6:5 110:1 327:344:10 45:15,21 46:3,4,16 50:8sudden 27 Suffice 37:166:7 167:20329:21 350:451:9,17 58:10,12sufficient 1000000000000000000000000000000000000	
162:10     164:8,12     110:1     327:3     46:3,4,16     50:8     Suffice     37:3	0:16 42:8 48:5 51:18
179:5 228:14 <b>STEMI</b> 232:21 59:10,12 61:20 286:5	122:3 129:13
241:7,18 285:13 step 59:11 301:13 63:17,22 64:3,9 suggest 22	
285:20 340:1 steps 5:4 344:10,11 64:15,20 65:5,6 231:22 2	
342:12 344:12 348:4 65:19 74:11 78:21 284:1	181:3 213:2
stated 240:3 316:2 349:10 79:3,5 83:2,10 suggested 3	
states 19:13 20:1 STETTLER 2:6 90:22 93:2 94:4 339:18	274:17 275:16,18
30:5 34:18 45:14 <b>STEVEN</b> 1:18 96:15 98:15 242:1 suggesting	
79:22 132:16,18 stick 32:17 263:12 210:17	295:4 298:4
19:22 192:10,10     stick 92:17     200:12     210:17       144:4,12 151:4     stir 306:20     STS's 58:20     suggestion	
144.4,12     151.4     still 500.20     510 5 50.20     suggestion       152:6,10,12     stop 7:5 11:9     stuck 184:2,2     suggests 13	
152:0;10;12     stop 7:5 11:5     stack 104:2;2     stagests 1:       155:13,17,21     303:15     studied 316:11     293:2	surgeon 17:19 18:8
155:15,17,21 505:15 studied 510:11 255:2 156:1,4 159:13,20 storm 306:20 studies 132:8 suitability	e
150:1,4     159:13,20     storm 500:20     studies 152.8     studies 152.1       163:6     166:7     straightforward     170:22     171:12,14     57:1     96:1	e
<b>U</b>	9 190:5,8 20:5 34:16 45:11
179.22     182.7     238.18     509.22     174.2     128.19,1       248:18,22     323:21     strategies     253:22     study     136:16     173:9     211:7,10	,
323:22 305:20 175:5 253:21 264:18 22	,
statistic 71:17 strategy 235:17 293:1 315:8 288:3,6 3	· · · · · · · · · · · · · · · · · · ·
statistic /1.1/     strategy 255.1/     255.1 515.8     268.5,0 5       108:17 140:19     stratification 301:1     stuff 178:10 306:14     326:1	318:1
108.17     stratified 228:11     stuff 178.10     500.14     520.1       149:21     189:5     stratified 228:11     subgroups 218:22     suitable 12	
203:16 227:19 stream 83:4 84:1,5 subject 2:2 34:11 suite 120:2	8.
205.10 227.17 stream 85.4 84.1,5 subject 2.2 54.11 suite 120.2 275:2,15 streamlined 10:8 50:14 173:7 summarize	
statistical 25:15 10:12 314:16 331:14	102:3 121:2
Statistical 25:15     10:12     514:10     551:14       28:11 124:14     Street 1:9     subjective 173:8     summarize	
$147\cdot14$ 148.16 strength 19.21 submission 9.4 summary	
147:14     148:16     strength     19:21     submission     9:4     summary       227:18     231:1     157:21     142:21     274:16     17:11     17:11	13:9 308:21 318:1
227:18 231:1     157:21     142:21 274:16     17:11 20	13:9308:21 318:1:13 36:1348:12
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19:16	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:6
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19:statistically 294:11263:17 264:8submitted 24:21211:18 25:	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:636:6304:1 334:6
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 2statistician 18:2stressed 176:276:1,1 90:5 217:1249:15 2	13:9   308:21 318:1     :13 36:1   348:12     3:13   surgical 29:5 293:6     36:6   304:1 334:6     90:7   surprised 144:18
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 22statistician 18:2stressed 176:276:1,1 90:5 217:1249:15 22129:21stretch 183:18294:12324:22 32	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:636:6304:1 334:690:7surprised 144:1848:3,8228:16
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 2statistician 18:2stressed 176:276:1,1 90:5 217:1249:15 2129:21stretch 183:18294:12324:22 3statistics 84:11strikes 323:8subsequent 41:2super 163:	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:636:6304:1 334:690:7surprised 144:1848:3,8228:1611surveillance 181:18
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 2statistician 18:2stressed 176:276:1,1 90:5 217:1249:15 2129:21stretch 183:18294:12324:22 3statistics 84:11strikes 323:8subsequent 41:2super 163:274:19 300:16strong 21:1,5 45:15subsequently 41:2support 9:	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:636:6304:1 334:690:7surprised 144:1848:3,8228:1611surveillance 181:189 27:3242:15 286:14
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 22statistician 18:2stressed 176:276:1,1 90:5 217:1249:15 22129:21stretch 183:18294:12324:22 32statistics 84:11strikes 323:8subsequent 41:2support 9:274:19 300:16strong 21:1,5 45:15333:2028:1 108	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:636:6304:1 334:690:7surprised 144:1848:3,8228:1611surveillance 181:189 27:3242:15 286:14:14 194:1SurveyMonkey
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 22statistician 18:2stressed 176:276:1,1 90:5 217:1249:15 22129:21stretch 183:18294:12324:22 34statistics 84:11strikes 323:8subsequent 41:2super 163:274:19 300:16strong 21:1,5 45:15333:2028:1 108status 8:3 145:16104:13 108:14subset 23:6 26:16195:2,4 2	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:636:6304:1 334:690:7surprised 144:1848:3,8228:1611surveillance 181:189 27:3242:15 286:14:14 194:1SurveyMonkey223:20347:15
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 22statistician 18:2stressed 176:276:1,1 90:5 217:1249:15 22129:21stretch 183:18294:12324:22 3statistics 84:11strikes 323:8subsequent 41:2super 163:274:19 300:16strong 21:1,5 45:15333:2028:1 108status 8:3 145:16104:13 108:14subset 23:6 26:16195:2,4 22161:11 168:15195:1 224:2,740:2,4 72:3,9253:18 20	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:636:6304:1 334:690:7surprised 144:1848:3,8228:1611surveillance 181:189 27:3242:15 286:14:14 194:1SurveyMonkey223:20347:1564:5survive 31:9
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 22statistician 18:2stressed 176:276:1,1 90:5 217:1249:15 22129:21stretch 183:18294:12324:22 32statistics 84:11strikes 323:8subsequent 41:2support 9:301:2,8 302:772:12 75:10333:2028:1 108status 8:3 145:16104:13 108:14subset 23:6 26:16195:2,4 22161:11 168:15195:1 224:2,740:2,4 72:3,9253:18 22177:4,13 179:14270:20 292:8174:7 207:11266:6 310	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:636:6304:1 334:690:7surprised 144:1848:3,8228:1611surveillance 181:189 27:3242:15 286:14:14 194:1SurveyMonkey223:20347:1564:5survive 31:90:19Susannah 2:17
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 22statistician 18:2stressed 176:276:1,1 90:5 217:1249:15 22129:21stretch 183:18294:12324:22 34statistics 84:11strikes 323:8subsequent 41:2support 92:301:2,8 302:772:12 75:10333:2028:1 108status 8:3 145:16104:13 108:14subset 23:6 26:16195:2,4 22161:11 168:15195:1 224:2,740:2,4 72:3,9253:18 24177:4,13 179:14270:20 292:8174:7 207:11266:6 314179:15 180:12,13stronger 273:20subsets 26:7support det	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:636:6304:1 334:690:7surprised 144:1848:3,8228:1611surveillance 181:189 27:3242:15 286:14:14 194:1SurveyMonkey223:20347:1564:5survive 31:90:19Susannah 2:1759:799:13 265:12
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 22statistician 18:2stressed 176:276:1,1 90:5 217:1249:15 22129:21stretch 183:18294:12324:22 3statistics 84:11strikes 323:8subsequent 41:2super 163:274:19 300:16strong 21:1,5 45:15subsequent 41:2support 9:301:2,8 302:772:12 75:10333:2028:1 108status 8:3 145:16104:13 108:14subset 23:6 26:16195:2,4 22161:11 168:15195:1 224:2,740:2,4 72:3,9253:18 24177:4,13 179:14270:20 292:8174:7 207:11266:6 314179:15 180:12,13stronger 273:20subsets 26:7support 9:210:12,13,14strongest 318:18substance 336:19131:15 24	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:636:6304:1 334:690:7surprised 144:1848:3,8228:1611surveillance 181:189 27:3242:15 286:14:14 194:1SurveyMonkey223:20347:1564:5survive 31:90:19Susannah 2:1759:799:13 265:1294:6286:7 329:18
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 22129:21stressed 176:276:1,1 90:5 217:1249:15 22129:21stretch 183:18294:12324:22 32statistics 84:11strikes 323:8subsequently 41:2support 9:301:2,8 302:772:12 75:10333:2028:1 108status 8:3 145:16104:13 108:14subset 23:6 26:16195:2,4 22161:11 168:15195:1 224:2,740:2,4 72:3,9253:18 24177:4,13 179:14270:20 292:8174:7 207:11266:6 314179:15 180:12,13stronger 273:20subset 23:6:19131:15 24210:12,13,14strongest 318:18substance 336:19131:15 24232:12 338:10,22strongly 210:3substantial 23:18311:18	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:636:6304:1 334:690:7surprised 144:1848:3,8228:1611surveillance 181:189 27:3242:15 286:14:14 194:1SurveyMonkey223:20347:1564:5survive 31:90:19Susannah 2:1759:799:13 265:1294:6286:7 329:18331:2
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 22129:21stressed 176:276:1,1 90:5 217:1249:15 22statistics 84:11strikes 323:8subsequent 41:2support 9:301:2,8 302:772:12 75:10333:2028:1 108status 8:3 145:16104:13 108:14subset 23:6 26:16195:2,4 2161:11 168:15195:1 224:2,740:2,4 72:3,9253:18 2177:4,13 179:14270:20 292:8174:7 207:11266:6 310179:15 180:12,13stronger 273:20subsets 26:7support def: 339:2,10 341:9,17232:12 338:10,22strongly 210:339:10311:1839:10suppose 32	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:636:6304:1 334:690:7surprised 144:1848:3,8228:1611surveillance 181:189 27:3242:15 286:14:14 194:1SurveyMonkey223:20347:1564:5survive 31:90:19Susannah 2:1759:799:13 265:1294:6286:7 329:18331:2Suter 3:6 57:17,17
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 2statistician 18:2stressed 176:276:1,1 90:5 217:1249:15 2129:21stretch 183:18294:12324:22 3statistics 84:11strikes 323:8subsequent 41:2support 9:301:2,8 302:772:12 75:10333:2028:1 108status 8:3 145:16104:13 108:14subset 23:6 26:16195:2,4 2161:11 168:15195:1 224:2,740:2,4 72:3,9253:18 2177:4,13 179:14270:20 292:8174:7 207:11266:6 31179:15 180:12,13stronger 273:20subsets 26:7support d232:12 338:10,22strongly 210:3substantial 23:18311:18339:2,10 341:9,17213:4 317:1739:10suppose 32stay 15:1 98:21struggle 262:3substantially 30:3supposed 12	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:636:6304:1 334:690:7surprised 144:1848:3,8228:1611surveillance 181:189 27:3242:15 286:14:14 194:1SurveyMonkey223:20347:1564:5survive 31:90:19Susannah 2:1759:799:13 265:1294:6286:7 329:18331:2Suter 3:6 57:17,1713:1458:1,6,9 64:1,1,22
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 2statistician 18:2stressed 176:276:1,1 90:5 217:1249:15 2129:21stretch 183:18294:12324:22 3statistics 84:11strikes 323:8subsequent 41:2support 9:301:2,8 302:772:12 75:10333:2028:1 108status 8:3 145:16104:13 108:14subset 23:6 26:16195:2,4 2161:11 168:15195:1 224:2,740:2,4 72:3,9253:18 2177:4,13 179:14270:20 292:8174:7 207:11266:6 314179:15 180:12,13stronger 273:20substance 336:19131:15 2232:12 338:10,22strongly 210:3substantial 23:18311:18339:2,10 341:9,17213:4 317:1739:10suppose 32stay 15:1 98:21struggling 285:4success 58:1634:5 113	13:9308:21 318:1:13 36:1348:123:13surgical 29:5 293:636:6304:1 334:690:7surprised 144:1848:3,8228:1611surveillance 181:189 27:3242:15 286:14:14 194:1SurveyMonkey223:20347:1564:5survive 31:90:19Susannah 2:1759:799:13 265:1294:6286:7 329:18331:2Suter 3:6 57:17,1713:1458:1,6,9 64:1,1,22:8 124:1865:6 67:16,22
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 22statistician 18:2stressed 176:276:1,1 90:5 217:1249:15 2129:21stretch 183:18294:12324:22 3statistics 84:11strikes 323:8subsequent 41:2support 9:301:2,8 302:772:12 75:10333:2028:1 108status 8:3 145:16104:13 108:14subset 23:6 26:16195:2,4 2161:11 168:15195:1 224:2,740:2,4 72:3,9253:18 2177:4,13 179:14270:20 292:8174:7 207:11266:6 314179:15 180:12,13stronger 273:20substance 336:19131:15 24232:12 338:10,22strongly 210:3substantial 23:18311:1839:2,10 341:9,17213:4 317:1739:10suppose 32stay 15:1 98:21struggling 285:4success 58:1634:5 113119:10,14 310:2STS 13:5 18:4successful 6:9326:14	13:9 $308:21\ 318:1$ :13 36:1 $348:12$ $3:13$ $surgical\ 29:5\ 293:6$ $36:6$ $304:1\ 334:6$ $90:7$ $surprised\ 144:18$ $48:3,8$ $228:16$ $11$ $surveillance\ 181:18$ $9\ 27:3$ $242:15\ 286:14$ $242:15\ 286:14$ $SurveyMonkey$ $223:20$ $347:15$ $64:5$ $survive\ 31:9$ $0:19$ $Susannah\ 2:17$ $99:13\ 265:12$ $286:7\ 329:18$ $94:6$ $286:7\ 329:18$ $22:16$ $Suter\ 3:6\ 57:17,17$ $13:14$ $58:1,6,9\ 64:1,1,22$ $:8\ 124:18$ $65:6\ 67:16,22$ $73:3\ 79:14\ 81:21$
227:18 231:1157:21142:21 274:1617:11 20283:3 301:16strengths 19:16submits 27:1095:13 19statistically 294:11263:17 264:8submitted 24:21211:18 2statistician 18:2stressed 176:276:1,1 90:5 217:1249:15 2129:21stretch 183:18294:12324:22 3statistics 84:11strikes 323:8subsequent 41:2support 9:301:2,8 302:772:12 75:10333:2028:1 108status 8:3 145:16104:13 108:14subset 23:6 26:16195:2,4 2161:11 168:15195:1 224:2,740:2,4 72:3,9253:18 2177:4,13 179:14270:20 292:8174:7 207:11266:6 314179:15 180:12,13stronger 273:20substance 336:19131:15 2232:12 338:10,22strongly 210:3substantial 23:18311:18339:2,10 341:9,17213:4 317:1739:10suppose 32stay 15:1 98:21struggling 285:4success 58:1634:5 113	13:9 $308:21\ 318:1$ $:13\ 36:1$ $348:12$ $3:13$ $surgical\ 29:5\ 293:6$ $36:6$ $304:1\ 334:6$ $90:7$ $surprised\ 144:18$ $48:3,8$ $228:16$ $11$ $surveillance\ 181:18$ $9\ 27:3$ $242:15\ 286:14$ $:14\ 194:1$ $surveyMonkey$ $223:20$ $347:15$ $64:5$ $survive\ 31:9$ $0:19$ $Susannah\ 2:17$ $99:13\ 265:12$ $94:6$ $286:7\ 329:18$ $331:2$ $22:16$ $Suter\ 3:6\ 57:17,17$ $13:14$ $58:1,6,9\ 64:1,1,22$ $:8\ 124:18$ $65:6\ 67:16,22$ $73:3\ 79:14\ 81:21$ $13:11$ $84:9,10\ 88:11,12$

suturing 102:3	331:12 334:21	tell 81:11,14,20	144:16 157:15	193:10,12 194:10
swaths 163:12	344:11	120:8 152:15	162:11 166:8	196:1,12,22 197:8
sweep 260:10	talked 48:14 49:7	171:14 241:1	188:12 232:17	198:1 201:2
sympathize 285:10	49:14 58:21 72:3	249:17	298:3 311:9	202:13 204:20
symptom 204:9	97:8 109:17	telling 81:8	323:18 341:3	205:17,22 206:10
symptoms 292:6	147:19 173:10	<b>ten</b> 272:4	test/retest 67:7	207:8 208:12,21
310:17	177:16 251:13	tend 154:7,8,9,11	273:3	211:14,17 212:20
syncope 292:15	257:16 298:2	186:17 219:14	tested 167:8 184:18	220:2,3 223:15
synergistic 84:19	321:19	307:17	184:20,22 185:6	255:20 260:19
synergy 11:19	talking 8:5 17:14	tended 142:3	185:11 193:1	263:11 264:9,22
system 1:15,15,16	32:9 103:2 118:16	tendency 117:20	198:5 218:22	270:10 272:6,21
1:17 34:11 50:20	118:18 149:16	tends 149:4 151:8	274:21 276:4	274:4,13 277:9
134:2 141:7 142:9	195:5 203:19	241:16	testing 107:17	279:15 280:5,17
252:6 308:20	235:4,5 252:4	tensions 126:13,17	162:9 166:5	281:22 287:7
systematically	254:3 268:15	<b>term</b> 151:19	178:21,21 179:5	288:11,12 294:22
254:9	323:11 340:21	terms 20:22 21:20	179:19 184:20	298:18 316:5
<b>Systems</b> 3:1 216:18	talks 193:12	22:6 23:3,11	185:20 186:2	318:11 325:14
	target 16:4 23:6	24:10,14 26:18	200:11 217:5	326:6,16 328:11
T	163:9 166:14	35:21 59:22 78:10	218:20 227:17	330:7 331:6
table 7:7 13:5	targeted 66:14	79:6 81:16 83:1	thank 6:6 11:15,16	339:21 340:10
57:11,13 98:22	334:18	87:22 88:18 96:16	16:17 20:17 21:13	343:21 344:2,4
103:1 221:4	Taroon 2:10 6:13	98:12 107:17	24:5 29:18 32:7	350:18 351:3,7,13
264:16 265:2	10:17 12:3 13:2	122:13 126:16	33:11 35:15 43:2	352:10
288:15 314:3	111:16 164:3	136:19 140:10,18	46:19 47:7 49:2	thanks 6:13,14
326:19 345:8	168:18 175:17,18	140:19 141:5	52:15 53:21 56:11	12:2 15:10,10
346:9	178:15 184:2	143:12,15 147:14	57:4,8,12 58:7	41:6 57:22 126:9
tabled 285:3	185:15 349:12	148:1,15 155:14	61:13,15,21,22	182:17 279:17
tad 57:22	350:16	175:12 181:21	73:3 82:8 84:9	286:19 314:15
tail 245:11 252:21	task 250:16 298:1	183:1 188:13,17	89:12 95:17 98:19	331:7
tails 87:15 94:15	<b>team</b> 35:8 99:1,13	199:4 203:12	102:21 106:19	<b>theater</b> 30:12 31:2
261:7,11	99:16 100:12	205:12 215:18	107:22 110:9	35:3
take 7:12 9:5 53:18	265:13 278:7,11	217:3,20,21	111:1 115:21	<b>theory</b> 282:21
158:18 164:15	278:21 281:17	218:20 223:20	119:19 120:6	therapy 320:4
234:16 254:7	326:19 350:18	224:21 225:19	121:6 126:10	thing 9:19 25:12
278:8 329:22	<b>teams</b> 265:1	226:18 227:1,16	128:10 129:1,7,16	81:9,12 85:17
336:3 339:9	teased 219:19	227:17 228:4	134:3,4,22 138:15	92:7 151:22 160:4
345:22 351:19	technical 76:1	230:13,19 235:19	139:18 141:12	180:15,21 181:13
352:3	100:13,13 108:13	238:19 244:16	145:5 146:19	207:20 239:22
<b>taken</b> 30:11 264:6	122:9 231:14	245:5 247:15	150:13 152:17	283:5 292:19
273:12 309:4	275:20 294:1	250:2 253:5	155:9 156:9 157:7	338:11 339:5
314:13	300:12 311:18	263:22 286:16	159:2 161:13	things 11:3 12:1
takes 9:14 48:1	319:21	287:17 312:21	165:3,16 168:8	15:12 30:17 49:13
talk 7:14 27:9	technique 319:18	348:3,3 349:10	170:4 175:15,16	70:12 85:20
65:14 75:18	techniques 228:2	351:6	178:14 180:5	101:13 111:21
112:13 114:14	319:20	terrain 290:1	186:15 187:21	115:2 121:14
138:16 193:14	teleconference 3:9	<b>terribly</b> 150:10	189:1,22 190:11	122:4 123:11
267:7 321:16	telephonic 314:18	test 42:19 132:13	190:12,22 193:9	145:16 150:5
	•	•	•	•

151:5 158:16	120:17 122:5,7,14	270:1,6,13,19	thoughts 94:1	thrombectomies
176:22 177:1	123:4,14 124:12	271:10,11 272:9	159:1 176:21	101:21
182:20 188:14	124:17,18,19	273:1,2,5 274:16	190:2 219:20	throw 253:3
194:3 209:17	125:5 127:10,11	274:21 275:5,6,8	249:2 255:22	<b>throwing</b> 236:15
210:10 235:7	128:11 132:9	275:20 276:12	thousand 76:12	277:22
254:13 257:21	133:22 134:12	280:22 281:11,15	threat 146:8 231:3	<b>tied</b> 316:1
261:1 263:21	135:21 143:19	281:17 282:4	232:4 234:14,20	<b>tight</b> 260:8
274:17 283:6,7	145:10 146:1,7	283:19,21 284:15	240:5	<b>tighter</b> 85:21
284:12 292:12,13	147:6 150:9	285:10 287:10,14	threats 141:4	tightest 342:20
292:13,17 301:7	154:18 155:3	288:20 289:10	145:10 228:4	<b>time</b> 9:6,13 10:18
306:15,16 308:6	157:11,20 158:1	292:16 293:10	235:8	15:17,17 17:3,8
309:8 310:17	158:11,15,17	294:7 295:11	three 23:5 25:4	21:11 22:16 24:1
311:1,2 312:11	163:4,15 164:7,13	296:2 298:5 299:2	26:19 47:1 49:20	26:21 30:20 31:17
316:14 317:5	164:17 165:6	299:11 300:13	49:22 54:7,11	33:7 40:17 41:1,3
318:7 333:2	167:14 168:21	307:6,11 309:8	66:19 69:3 86:5	41:4 43:22 47:2
341:13 342:2,4	169:10,14 170:8	312:8 314:3 315:1	86:19 89:6 95:8	47:17 56:5,21
think 8:3,10 10:7	170:12 171:1,3,20	315:6,22 316:2,10	106:6,22 108:4	62:17 66:6,20
11:20 12:10 19:14	172:15 173:6	316:17,22 317:16	110:4,18 126:12	71:9 77:11 86:6
19:18 20:13,15,21	177:2,5,10,13,20	317:21 318:7	128:4 129:4 137:8	86:20 91:22 95:5
21:3,4,11 23:9	178:7,11,11	324:3 325:3,7,9	138:10 139:12,16	96:2 97:1 104:18
24:8,10,18 26:11	181:12 182:21	325:10 347:7	157:2,6 169:21	106:6 107:1 108:5
26:19 30:1 32:8	183:1,2,9 185:6	thinking 115:5	170:2 183:5	110:5,19 124:5
34:7,12 35:16	185:17 187:14	235:19 246:5	189:15 196:17,20	125:4,4,16 126:13
38:6,17 39:15	189:11 193:6,11	252:4 304:12	197:17 202:6	127:22 128:5,16
41:3 42:11 44:12	199:2 208:3	310:15 316:13	206:4 207:2 209:2	128:17 129:6
46:12 47:18,22	209:10 212:16	third 101:1,22	220:20 225:13	134:18 137:8
48:9 49:19 50:6	213:10 214:2	346:4	226:10,12,14	138:11 139:13
50:11,17 51:3,8	215:14,16 218:21	thirds 292:3	227:9 240:9 243:3	144:10 157:3
51:17,22 52:6,11	219:3 220:1,9	thirteen 56:8 66:8	249:5,7,11,12	168:17 169:22
53:12 54:15,21	223:13 224:7	138:13 243:7	272:1,15 274:7,11	180:6 183:7
55:10 58:3,9,15	230:14,18 232:3,8	Thirty 54:11	276:1 279:21	188:19 189:16
58:21 59:14,21	233:13 234:5,10	<b>THOMAS</b> 2:5	280:12 287:1	190:6 194:14
60:14,18 62:9	235:7 237:7,13	thoracic 2:19 3:3	292:3 297:15	196:17 197:18
63:11,20 68:5	238:6,15,17,21	4:5 18:7,9,19 19:9	298:13 300:3	198:14 202:7
72:12,19 74:1	239:2 240:2	20:5 34:16 45:11	310:16 317:21	206:5 207:3 209:3
76:3 77:22 78:7,9	245:12 246:10	67:12 76:6 102:15	320:15 321:6	211:8 212:5,19
79:14 80:4,15	247:2,6,8,11	thought 35:22 44:7	322:4 324:16	215:2,20 220:5
82:6 83:13,19	248:2,9,12 250:8	83:17 115:12	331:12 344:19	223:13 224:16
84:9,13 86:14	250:12,19 251:5	141:19 158:12	three-year 267:18	225:13 226:11
87:4,6,12,18 88:1	251:18 252:1,3,15	173:14 184:12	273:15,21	227:10 230:15
92:4,16 93:8,9,12	253:4,7,13 254:2	204:8 257:11,12	threshold 26:1,14	232:14,22 239:6
94:5,7,21 97:22	254:11 256:1,5,11	264:7 296:9	250:13 261:18	240:10 242:5,10
98:2,6,10,11,17	257:7,19 259:1	302:21 311:14	317:1,10	243:4 244:9 247:1
105:19 106:1	260:6,16 261:11	313:13 314:22	thresholds 26:5	248:18 249:6,9
111:9,22 112:12	262:1,3,4,7,9,10	thoughtful 302:17	257:8	251:9 255:15
112:15 113:21	263:6,15,18 264:2	thoughtfulness	thrilled 351:15,19	260:17 264:13
115:7,12,18 118:7	264:4,6 269:12	6:19	throated 13:1	266:17 269:17
I		I	I	1

271:3 272:1,16	328:20 330:12	trouble 122:4	224:18	298:13 300:3,16
274:8 275:9,22	335:10	148:11 261:11	twenty-one 71:11	301:7 310:15
279:22 280:13	total 64:19 149:22	troubleshooting	134:21	320:15 321:6
287:2 288:4 296:4	245:11,13 301:17	157:19	Twenty-two 62:20	324:16 325:17
296:18 297:15	totally 72:8 254:21	<b>true</b> 25:14 30:22	<b>twice</b> 94:14	331:7 332:14
298:14 300:4	285:3	69:5 85:8 159:6	twist 288:21 289:20	335:5,10,15,20
304:3 305:16	touch 169:11	160:18 166:2	<b>two</b> 10:8 15:5,6,11	344:6 345:4,6
306:13 310:3	touched 36:19	170:19 174:1	16:2,7 18:6,17	346:9 351:1,5
311:14 315:11	182:5 340:1	175:3 177:20	19:2,4,6,15 20:14	<b>two-minute</b> 17:11
318:20 320:16	tough 181:13	192:17 239:7	23:5 42:3 45:17	<b>type</b> 25:17 75:7
321:7 324:17	217:12	268:2 312:20	46:15 47:1 48:8	126:7 273:5 275:6
325:12,18,21	track 12:14 142:3	truly 35:2 183:3	48:12 49:21 58:12	309:1 310:12
327:14 331:11	200:13 216:8	truncate 16:19	58:14 66:19 68:9	types 142:3 154:4
333:1 334:20	tracked 31:19	try 15:13,20 82:2	69:18,20 76:2	228:8
350:14 351:4	tracks 190:14	85:3 114:9 127:14	78:1 81:8,11	typical 102:7 199:4
352:7	201:21	173:12 208:14	84:10 86:5,9,19	287:15
<b>Timeline</b> 5:4	tract 195:12 197:11	244:14 254:9	87:1 92:5,19	typically 200:3
timely 12:22	201:16	268:17 297:17	93:10,10 94:10	229:17 231:7
159:17	tradeoff 34:7 80:18	305:15 338:13	97:8 104:18,19	232:18 304:17
times 135:18	tradeoffs 126:18	trying 65:16 68:4	106:5,22 108:4,8	
256:18 284:22	transfer 304:2	78:12 97:4 112:21	110:4,18 111:20	U
<b>tiny</b> 337:12	transferred 38:3	115:5 117:14	128:4,15 129:14	<b>U.S</b> 78:20 241:1
tip 277:15	transfers 120:9	125:9 142:10	134:18 135:13,16	282:9 297:22
<b>tired</b> 279:1	transitions 174:19	144:5 146:5,21	137:7 138:10,14	<b>ultimate</b> 123:14
today 11:8 12:7	194:5	154:1 158:13,14	139:12,17 146:9	ultimately 65:15
15:4,12,18,22	transparency	159:13 163:4,16	149:16 150:2,9,16	101:5 141:21
16:8 17:21 18:16	46:18	166:16,20 170:21	152:20 157:2	unable 28:22 81:21
70:4 103:11 130:4	transparent 61:13	180:15 181:13	158:16 169:21	<b>unaware</b> 64:10
209:16 221:12,16	trauma 151:15	185:7 217:10	176:20 189:15	292:21 293:4
222:2 265:14	<b>TRAVIS</b> 2:7	230:18 232:22	190:5 191:6	uncertainty 52:10
284:21 285:18	184:15 256:16	244:11 251:6	194:14,16 196:5,8	77:8 282:18
347:1 348:9	295:6 296:7 297:2	254:6 261:16	196:16 197:17	uncomfortable
350:15	297:22 321:1,15	308:2 310:13	202:6,8 206:4	314:9
today's 12:22 97:1	<b>treat</b> 319:10	316:8 327:4	207:2,4 209:2	undergo 29:10
told 323:6	treated 217:18	337:21	210:1 211:7	undergoes 30:12
<b>Tom</b> 161:15	treatment 336:9	TUESDAY 1:5	220:22 224:16	undergoing 34:19
tongue 277:16	treatments 341:16	<b>turn</b> 13:7 16:12	225:12 226:10	320:10
<b>Tony</b> 150:15	tremendous 28:10	42:14 51:22 62:2	227:5,9 229:1	undergone 42:3
155:10	90:20	103:7 211:20	240:9,13 243:3	underlie 158:20
tool 134:1 193:7	trends 247:9	327:18 344:10	249:5 251:3	underlying 85:6,10
301:1	trial 8:19	348:2 350:16	262:21 264:6,12	149:10
toolkit 254:15	trials 281:19	turned 335:3	264:13 268:4	understand 28:19
top 76:21 188:21	triangulate 305:10	turning 326:9,10	271:3,22 272:15	30:1 41:17 68:4
292:11	tried 30:15 160:4	<b>Twelve</b> 197:21	274:7 276:20	76:4 78:18 81:16
<b>topic</b> 52:3 82:10	247:17 305:10	twenty 86:9 87:1	279:21 280:12	82:2 85:3 92:10
240:1 328:9 347:8	339:8 341:20	104:21 110:7	287:1 288:3 292:2	92:15 97:19 99:17
topics 288:22	342:17	137:11 206:8	296:18 297:14	123:17 153:14
		10,111 200.0		l

	1			
238:5,11 257:17	32:4 34:8 43:8	189:13,13,14,19	127:12 159:22	145:11 146:8,22
257:22 293:4,7	50:22 59:4 74:14	207:9,10,13 209:1	193:7 255:18	156:22 157:2,5
302:5 319:12	96:5 107:10 132:7	209:4 214:21	user 164:19 256:9	184:19 201:4
320:2 327:5	139:21 140:2	243:9 246:2 249:3	313:10	202:14,16,17,21
337:21	170:10 171:5	249:4,11 250:17	users 60:22 61:13	203:4,12 204:22
understandable	172:11,14,20	250:18 251:20	uses 80:13 83:3	205:3,18,20 206:2
125:17	173:2,19 203:13	280:18 282:2	89:17 133:16	206:3,8 213:20
understanding	207:18 217:9	285:5 286:21,22	148:4 149:13,13	226:17 227:15,16
39:16 41:21 48:15	222:7,11,15	287:4 321:12	159:7 267:9	227:20 228:4
94:5 98:13 158:8	262:13 266:13	324:13,15,19	284:18,18 294:8	231:3 232:4
186:6 209:21	267:9 290:10,22	usable 127:15	294:10	234:21 235:8,19
234:1 235:11	295:21 327:1	253:12,13 256:14	usual 266:19	235:22 236:22
253:21 301:17	333:18 334:9	256:14	usually 68:19 85:17	240:7,9,13 263:9
understands	348:10	<b>usage</b> 30:10	149:14 185:3	274:14 279:21
275:16	unrelated 218:2	<b>use</b> 8:19 14:17 20:9	231:7 241:17	280:3 300:10,11
understatement	unscheduled	30:6 35:18,20	utilization 119:3	300:13 320:13,15
23:10	181:22 313:4	45:5 47:8 55:5	utilized 51:19	320:18
underway 173:10	unspecified 102:16	56:4,8 59:6 60:22	<b>UW</b> 2:1	valuable 91:8 134:1
281:8	upcoming 284:16	69:3 77:5 87:16		259:2
underwent 28:17	update 221:3	88:19 95:4,8	V	value 231:20
undocumented	235:20 246:5	97:17,21 111:2,10	<b>VA</b> 268:6,9,10	233:16 239:11,12
343:13	250:16 327:19	112:2 115:5 123:6	<b>VAD</b> 28:20 29:16	246:8 253:7 315:2
unexpected 313:12	329:4,11 330:4	123:15,19 127:11	50:22 74:6,7,14	315:6 316:3
<b>Unfortunately</b> 82:7	344:3	128:4,8 130:8	74:16,19 75:1	values 157:18
<b>uniform</b> 144:6	updated 222:12,19	132:3 140:11	<b>VADs</b> 43:7 75:20	228:12 231:18,18
uniformly 199:10	236:4 274:17	142:22 144:7	valid 85:22 253:11	304:5,5
unintended 35:19	332:22 334:14	147:2,16 148:16	validate 303:9	Vancouver 130:4
48:6 49:21 74:3	updates 235:13	160:13 162:3	validated 108:16	variability 208:8
216:10 247:21	328:18 330:11	163:6 165:21,22	165:19 166:1	297:3 323:7,10
248:4 251:15	updating 236:20	166:6 169:5 172:5	301:12	337:22
270:2 286:13,16	<b>upper</b> 107:20	180:4,14 186:14	validating 79:16	variable 70:18
322:2 330:2	108:18	189:15,19 209:2,5	validation 58:18	142:14 230:4
<b>unique</b> 51:20 84:17	<b>urge</b> 93:1	210:13 214:20,22	64:13 76:4,13	233:22 234:9,12
<b>unit</b> 167:10	urgency 182:7	222:20 228:15	78:17 79:15 140:6	234:19 237:10
<b>United</b> 2:6 19:13	233:4	238:22 249:5,11	213:18 301:9,13	239:5,8,9 302:17
19:22 30:5 34:18	urgent 229:18	250:16,19,22	333:21	variables 20:9,11
45:13 79:22	<b>usability</b> 47:8,10	251:10,16,20	<b>validity</b> 32:8,16	37:17 82:17 84:4
<b>units</b> 337:5	48:20 56:2,3,8	256:9 262:11	33:12 38:7,16	85:7,8,17 144:13
University 1:14,15	87:3,5 92:4 95:2,3	277:5 280:18	41:8 43:4,21 44:3	199:17 204:5,9
1:18,19,21 2:17	95:8 111:2,7,10	282:3 286:21	48:20 59:21 67:4	235:15 239:10
2:18,18,19,20 3:2	123:6,15 124:18	287:1,5 301:21	71:13,16 72:11	335:18
3:2,3,8 18:3 25:11	125:6 126:16	302:19 318:7	81:4,7,13,17,18	<b>variance</b> 70:20
<b>unknown</b> 302:2	128:2,3,8 161:14	321:16 324:16,20	81:20 82:11 85:19	135:11
unnecessary 319:1	161:18 162:14	340:18 344:15	86:3,5,9 108:10	variation 25:14,15
<b>unplanned</b> 4:6 5:1	169:15 170:5,6	345:5,15	108:15 109:3	26:4 63:15,21
13:16,19 18:21	175:20 176:21	useful 39:13 80:6	110:2,4,7 139:19	65:17,22 100:5
30:11,20 31:8,20	178:11 186:11	94:7 121:17 123:9	140:11 141:5	159:4 224:1,3
	1			1

291:16,18 327:9virtually 55:6224:14 227:11325:22112:5 115:varied 155:22virtue 245:16249:15 264:10voting 22:14 23:21121:12 122variety 30:14visit 4:21 288:19271:1,20 272:2,1333:5 43:20 46:22123:5 127:337:20311:10 313:11273:8 274:556:3,18,19 62:16142:19 148various 332:17316:15 317:19279:19 280:1066:4,18 71:7 86:4153:14 154333:10326:3 349:8286:21 287:1886:18 95:3,22169:18 170vary 157:13visits 34:19 290:10288:1 296:16,19104:17 106:4,20172:6 181:	2:1 22 3:15 4:21 0:7 10 3:14 3:10 2:20
variety 30:14visit 4:21 288:19271:1,20 272:2,1333:5 43:20 46:22123:5 127:337:20311:10 313:11273:8 274:556:3,18,19 62:16142:19 148various 332:17316:15 317:19279:19 280:1066:4,18 71:7 86:4153:14 154333:10326:3 349:8286:21 287:1886:18 95:3,22169:18 176	22 3:15 4:21 0:7 10 3:14 3:10 2:20
337:20311:10 313:11273:8 274:556:3,18,19 62:16142:19 148various 332:17316:15 317:19279:19 280:1066:4,18 71:7 86:4153:14 154333:10326:3 349:8286:21 287:1886:18 95:3,22169:18 170	3:15 4:21 0:7 10 3:14 3:10 2:20
various 332:17316:15 317:19279:19 280:1066:4,18 71:7 86:4153:14 154333:10326:3 349:8286:21 287:1886:18 95:3,22169:18 170	4:21 ):7 10 3:14 3:10 2:20
333:10     326:3 349:8     286:21 287:18     86:18 95:3,22     169:18 170	):7 10 3:14 3:10 2:20
, , , , , , , , , , , , , , , , , , , ,	10 3:14 3:10 2:20
vary 157:13 visits 34:19 290:10 288:1 296:16,19 104:17 106:4,20 172:6 181:	3:14 3:10 2:20
	8:10 2:20
varying 327:7     290:22 291:1,21     297:12 300:5     106:21 108:3     185:15 193	2:20
vascular 4:10 15:6     293:3 295:21     325:15 328:13     110:1,3,17 125:6     204:21 208	
99:5,22 100:17 301:18 304:11 335:13 347:10 128:3,14 134:17 209:22 212	229:10
101:16 102:2     305:17 313:15     350:9     137:6 139:11     214:15,21	
113:14 114:21   314:12 315:4   voted 24:3 44:3,3   157:1 169:20   229:12 232	
115:3,4 116:1   319:8   47:5 66:8,9,22   189:14 190:4   237:21 25 1	
126:22 127:3,12     visually 77:1     67:1 71:11 87:1     194:13 196:15     251:12 25:11	
128:21 156:13     volume 48:12 49:12     107:4,4 110:7,7     197:16 202:5     257:17 265	
348:21     55:14 69:19     110:21,21 157:5,5     207:1 209:1 211:6     270:11 27:5	
<b>vast</b> 237:7 311:6 138:22 148:1 170:2 194:16,16 224:15 225:11 285:14,15,	
vastly 283:17     153:4,12,19 155:2     196:20 197:21     226:9 227:8 240:8     286:9 295:	
venal 101:12     156:6 245:6     206:8 274:11     243:2 249:4     306:19,20,	
ventilator-depen     267:20 299:19     335:8     264:11 271:2,21     311:21 319	
219:13     303:6 307:22     votes 21:12 22:18     272:14 274:6     328:19 330	
ventricular 28:18     308:2,12 318:10     24:2 33:8 44:2     279:20 280:11     332:20 344	
29:1,10,14 30:6     volumes 26:17     47:4 56:7,22     286:22 287:21     351:12 352	
30:10,19 31:2,7   148:10   62:19 66:7,21   288:2 296:17   wanted 6:10	
31:11,16,18 32:4     voluntarily 256:22     71:10 86:8,22     297:13 298:12     15:11,19 1	
32:6 33:18 34:8     258:3 260:5     95:7 96:3 104:19     300:2 320:14     73:4 75:19	
34:10 35:4,7     voluntary 223:7     104:20 107:3     321:5 324:15     99:18 102:	,
version 53:1 222:14     256:19 257:2,13     108:7 110:6,20     325:1,16 347:15     102:20 116	
279:10,11     volunteers 7:7     128:7,17,18,22     121:11 160       versus 26:19 50:22     15:16     134:20 137:10     W     164:10 179	
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	
171.1172.11 71.0 95.2,12 197.20 202.0,10 325.10	1:9
	.11
	9
	25.16
	1:9
	5
	13
Virginia 1:13     209:8 211:4     321:9 324:19     109:15 111:17     waste 10:20	

		1	1	
way 8:9,13 9:12	17:6,9 21:9 22:6	we've 20:15 32:18	Wes 63:8 65:10	290:2 307:19
13:1 19:18 30:2	22:22 32:9 33:12	36:7,11,19 39:2,6	82:11 83:7,16	312:11 351:6,8
34:12 42:16 53:10	37:18 43:3,11,13	49:7,14 51:17	85:13 95:13,20	352:11
84:5 89:2,10,11	44:5 46:17,20	58:21 63:1 71:2	312:2 313:22	worked 39:5 41:11
93:14,18,19 97:21	47:8 48:6 54:15	103:1,6 105:14	340:11 345:3,10	workgroup 15:17
115:18 117:1	55:20 56:12 57:19	106:1 131:22	Wes's 94:9,9	58:20 68:1 72:17
118:7 123:16	62:1 65:12,15	137:1 147:19	<b>WESLEY</b> 1:18	75:9 103:14,16,20
124:15,16 127:10	80:22 81:18 83:8	153:8 154:6,19	West 1:13	104:2,5,6,15
132:4 136:1 140:1	83:9,13 89:11	160:6 161:3 165:8	whatnot 24:12	106:13 255:12
140:3 142:22	90:7 104:16	166:4 173:10	whim 315:19	350:21
149:5 150:5,11	112:12 113:16	177:16 185:4	whispering 103:2,7	working 61:20
155:3 159:17	114:9 118:15,17	199:2 202:2	white 210:20	109:22 113:5
160:6 164:7,13	124:17 125:20,21	206:16 207:12	284:16	125:9,20 131:5
178:12 184:17	127:21 128:2,11	209:15 212:18	wide 69:2 204:12	165:12,13 172:22
188:2 198:5 238:8	135:20 145:6,14	219:10 222:19	widely 110:12	224:9 253:16,17
242:14 246:4	145:18,21 146:12	228:20 238:8,8	147:2 172:17	255:13 293:14
253:1 254:13	149:12,16 150:4	239:14,21 240:3	301:22	324:11
282:7 283:1,4,7	161:14 162:14	242:11 246:10,20	wider 196:10	works 111:18
284:8 322:19	163:16 166:16	248:2,15,16	widespread 130:6	177:17 269:14
334:8 341:22	168:4 170:5	251:13,19 263:2	willing 256:22	318:9
ways 10:7,19 42:3	172:21 173:19	265:4 274:2 281:2	257:11	world 9:16,17
46:15 124:10	174:13 177:11,13	281:16 287:16	window 137:17	45:12 84:16 134:9
125:12 143:17	180:6,18,21 184:2	288:22 310:6	251:1 258:2 260:7	182:2 323:13
146:16 149:20	184:7,8 195:15	327:16 348:9	311:14	worried 336:12
158:19 167:2	201:22 202:8,20	351:1	within-state 160:3	worry 148:9
195:1 238:10	206:20 211:19	weaknesses 19:16	Women 2:4	worse 77:15 89:7
246:5 252:1 282:5	214:12,13 218:17	<b>web</b> 46:1	Women's 1:22	178:4 247:16
282:6 283:2,6,10	221:3 232:22	website 46:5,9,16	wondered 71:20	261:19 282:11,20
283:17 305:12	233:13 235:5	88:8 91:11 259:12	wonderful 351:16	323:5
332:17 338:12	236:13,14,17	websites 223:9	wondering 73:1	worst 248:5 261:8
342:17	244:11 251:6	week 315:20	150:19 151:3	worth 25:4 140:2
we'll 14:1 15:20	253:20 254:3,5,13	weeks 7:22 258:2	153:16 213:16	140:17 157:11
16:9 60:15 62:15	257:18 263:18	260:11	228:17 310:9,18	170:9 237:17
65:13 71:6 86:2	264:1 269:15	weight 7:2 282:17	word 34:2 116:3	worthwhile 351:21
138:16 215:16,19	270:1 273:14	<b>Weill</b> 347:9	282:1	wouldn't 40:13
220:15 221:4,6	282:1 289:21,22	welcome 4:2 6:4,10	words 118:3 236:11	57:15 117:12
224:13 233:1	295:1 298:2	6:11 13:4 90:1	266:12	121:12,20 136:18
265:1 270:8	299:17,17 302:11	265:3 326:18	work 6:7,15 7:2,10	150:10 160:1
275:12 288:14,17	304:15 305:16,19	347:8	7:14 11:11 12:9	181:18 182:2
288:22 289:11	307:2 308:5 317:9	went 28:12 35:3	15:14 80:21 116:7	wrestled 314:17,21
295:2 296:14	325:8,19 326:14	75:13 129:9,9	122:12 149:11	write 111:21
312:6 345:7 346:8	328:15 329:7	140:6 220:18,18	162:10 186:21	writing 349:17
346:10,12,20	335:15 336:12	257:22 271:11	212:22 217:5	written 34:2 94:3
347:13,17 350:4	337:14 339:14	279:11 352:14	244:14 247:2	94:19 97:9 208:2
350:13,13 352:3	343:6 344:20,22	weren't 13:22	268:11 273:13	208:8
<b>we're</b> 8:8 9:8 10:14	345:10,17,22	152:6 205:1	281:20 283:9	wrong 123:18
10:22 12:15 15:3	349:11,16 351:10	220:11	284:3,5 286:11,14	229:11 240:3
		1	I	I

250:7 259:22	16:2,14 96:22	107:20,21 108:18	104:6 120:1,1	323:18
322:22	182:5 209:16	146:7 227:4	143:10 202:11	<b>1030</b> 1:9
wrongfully 322:13	220:11 249:20	170.7 227.7	207:6 212:4	<b>1050</b> 1.9 <b>10th</b> 105:5 225:5
www.sts.org 46:5	vielded 68:17	0	216:16 279:10	<b>11.8</b> 236:9
www.sts.org+0.5	107:18	<b>0.001</b> 282:19	300:7 312:18,18	<b>11:41</b> 220:18
X	<b>York</b> 157:16	<b>0.03</b> 75:4	315:9,10 324:11	<b>12</b> 63:4 260:11
<b>X</b> 249:22	162:10 165:13	<b>0.33</b> 68:17	324:20,21 326:4	298:16
	166:9 179:5	<b>0.331</b> 67:20	352:7	<b>12.3</b> 105:6 236:8
Y	192:22	0.335 299:2	<b>1-year</b> 41:1 203:1	<b>12.5</b> 105.0 250.8 <b>12.5</b> 21:22
Yale 2:17,18,18,19	younger 42:4	<b>0.37</b> 227:2	<b>1</b> (a) 21:10,12 62:16	<b>12.3</b> 21.22 <b>12.6</b> 196:9
2:20 3:2,3,6,8 4:8	276:14	<b>0.38</b> 273:4	62:19 104:17,20	<b>12.0</b> 196:9 <b>12.9</b> 105:10
4:10,20,22 19:1,7	2/0:14	<b>0.4</b> 68:18 107:19	,	<b>12:9</b> 103:10 <b>12:13</b> 220:19
19:14,19,21 20:3	Z	<b>0.43</b> 299:9	134:17 194:13,15	
42:17 57:11,18	Zaslavsky 3:7	<b>0.47</b> 27:11 53:3	224:15 271:2,5 296:17,20	<b>12:15</b> 220:16 <b>13.5</b> 21:17 225:6
64:2,4 72:22	129:20,21 149:7,7	<b>0.48</b> 273:19	,	<b>13.5</b> 21:17 225:0 <b>131</b> 4:12
98:20 99:5,11,13	129.20,21 149.7,7	<b>0.5</b> 70:1 139:1	<b>1(b)</b> 22:14,18 66:4	
103:5 112:15	<b>ZEHRA</b> 2:15	198:7 201:13	66:7 106:4,8	<b>14</b> 128:22 298:6
148:4 150:11	<b>zero</b> 21:12 22:19,20	<b>0.58</b> 275:21	137:6,10 196:15	321:10
167:5 221:1,11,16	24:4,4 33:10,10	<b>0.6</b> 274:19	225:11 271:21	<b>14.3</b> 105:10
265:1,13,18	44:4 47:6,6 56:9,9	<b>0.63</b> 71:17	272:3 297:13,19	<b>14.9</b> 105:7
278:21 279:3	57:3 62:20 66:9,9	<b>0.64</b> 276:1	<b>1(c)</b> 23:21 24:2	<b>15</b> 22:1 98:14
288:13,15 289:12	67:1,1 71:12,12	<b>0.641</b> 27:12	66:18,21 106:21	143:12 219:15
289:15,17 326:19	86:9,10 87:2,2	<b>0.66</b> 227:20	107:3 138:9,12	226:2
331:1	95:9 106:10,10	<b>0.67</b> 108:18 227:20	197:16,20 226:9	<b>150,000</b> 74:15
year 9:11 25:4	107:5,5 108:9,9	276:18 300:17	226:13 272:14,18	<b>15th</b> 1:9
26:18,19 34:20	110:7,8,22,22	<b>0.69</b> 140:19	298:12,15	<b>16</b> 22:19 324:20
37:19 54:3,9	137:12 138:14	<b>0.7</b> 68:18	<b>1,000</b> 14:11,15	344:14 349:13
73:22 136:4 242:8	139:17 157:5,6	<b>0.71</b> 203:17	25:19 291:14,17	<b>16.2</b> 291:14 333:19
246:21 250:15	170:3 189:19,20	<b>0.77</b> 198:8	295:22	<b>16.5</b> 74:22
251:19 255:4	196:21,21 197:22	<b>0.78</b> 25:7	<b>1,200</b> 257:7	<b>16.8</b> 63:4 333:19
267:1,2 269:4	202:12 206:8,9	<b>0.9</b> 77:4 105:21	<b>1,743</b> 201:11	<b>163</b> 332:11
275:21 279:9	207:7 209:5,6	<b>0.998</b> 276:20	<b>1.15</b> 143:11	<b>16th</b> 345:12
year's 257:12	224:18 225:2,16	<b>0171</b> 345:15	1.2 90:5,7	<b>17</b> 21:21 190:10
years 25:4 54:7,11	225:17 226:14,15	<b>0173</b> 345:13	<b>1.4</b> 73:16	207:6 209:5
67:11 69:3 123:21	227:14,14 240:13	<b>0327</b> 14:22	<b>1.5</b> 90:6,7	288:10 300:7
126:12 175:5	240:14 243:7	<b>0505</b> 4:18 288:6	<b>1.7</b> 196:7	326:4
180:20 183:5	249:12 262:15	349:3	<b>10</b> 64:16 67:12 73:7	<b>17.3</b> 271:13
222:6,10 228:20	264:21 271:6	<b>0695</b> 4:16 264:19	105:11 144:20	<b>17.5</b> 74:20
242:6 245:3	272:5,5,19,20	348:22	219:15 226:2	<b>17.9</b> 271:12
246:15 252:11	274:12,12 280:3,4	<b>09</b> 67:12	231:8 233:7	<b>1768</b> 327:20
253:15 267:14	280:15,15 287:5	0907.12	271:11 302:14	<b>1789</b> 5:1 15:22
268:4 271:12	297:20,20 298:17	1	10.1 225:6	221:3 326:9,22
273:14 276:1	298:17 300:8,8	<b>1</b> 4:2 13:9,12 21:10	<b>10.3</b> 135:17	327:16
284:14 300:16	320:18,19,19	22:15 23:22 30:4	<b>10:03</b> 129:9	<b>18</b> 4:5 22:1 41:14
331:12	321:10,11 324:21	33:6 43:21 56:4	<b>10:13</b> 129:10	41:15 42:10,21
yesterday 6:7,17	<b>zeroing</b> 305:17	56:20 62:17 66:5	<b>10:15</b> 129:7	105:11 167:9
12:4 15:14,18	<b>zone</b> 32:19 70:22	71:8 95:4 96:1,8	<b>100</b> 19:13,22 26:8	202:11 211:12
,	Zone 32.17 10.22	,0	37:7 38:21 78:6	276:5,18 289:4

320:18 340:18	<b>200,000</b> 334:7	<b>26</b> 132:15 155:13	343:4	349:19
<b>18.3</b> 271:14	<b>2004</b> 298:7	<b>265</b> 4:20	<b>320</b> 257:10	<b>60</b> 79:11 80:3
<b>18.7</b> 279:12	<b>2005</b> 226:4	<b>27,000</b> 291:20	<b>326</b> 5:2	<b>607</b> 139:2
<b>19.0</b> 279:11	<b>2008</b> 67:11 266:17	<b>28</b> 225:3	<b>331</b> 343:4	<b>62</b> 201:15
<b>19.4</b> 271:12	<b>2008-2010</b> 88:13	<b>288</b> 4:22	<b>34</b> 295:20	<b>65</b> 26:10 36:4,15
<b>19.7</b> 279:11	<b>2009</b> 267:1	<b>29</b> 228:4 282:10	<b>34.2</b> 21:22	39:18 41:13 42:1
<b>191</b> 4:14	<b>2010</b> 298:21 301:11		<b>344</b> 5:4	42:5,6,20 60:18
	<b>2011</b> 301:12 326:22	3	<b>349</b> 5:6	130:20,21 158:3
2	331:6	<b>3</b> 22:2,15 23:22	<b>350</b> 257:8	168:19 222:10
<b>2</b> 6:4 13:10 21:10	<b>2012</b> 270:16 281:3	33:6 37:18 43:21	<b>365</b> 37:12	
22:15 23:22 33:6	331:10,17	46:22 56:4 66:5	<b>3M</b> 3:1 216:17	7
43:21 54:10,10	<b>2014</b> 1:6	71:8 86:18 95:4		<b>7</b> 136:12 290:22
56:4,20 62:17	2017 94:3,20	123:21 207:6	4	349:19
66:5 71:8 95:4	<b>21</b> 96:7	211:13 219:17	<b>4</b> 22:15 23:22 33:6	<b>7-day</b> 4:21 326:2
96:1 130:17	<b>211</b> 4:15	222:6 242:6 243:2	43:21 56:4 66:5	349:7
132:18 135:4	<b>22</b> 21:12,12 22:18	246:15 337:9	71:8 95:4 122:18	<b>7:57</b> 6:2
202:12 244:20	24:2 57:3	<b>3-year</b> 27:13 28:2	135:18 209:6	<b>72</b> 132:19
245:3,10,11,12	<b>22.1</b> 63:5	37:19,21 41:3,4	222:6 321:10	<b>75th</b> 136:6
246:6 279:11	<b>222</b> 4:17	47:22 68:15 69:17	344:15 349:13,21	0
288:10 289:9	<b>229</b> 201:11	69:18,19 156:15	<b>4,000</b> 154:15	8
344:15 349:13	<b>23</b> 282:9	213:3,17 214:8,9	<b>4,464</b> 282:8	<b>8</b> 258:2 291:16
<b>2(a)</b> 33:5,8 71:7,10	<b>2375</b> 14:4 346:1	215:20 216:12	<b>4,500</b> 218:14	<b>8,000</b> 324:4
108:3,7 139:11,15	<b>2380</b> 14:20	251:1 269:16	<b>4,652</b> 332:9	<b>8.2</b> 295:22
202:5,10 227:8,12	<b>2393</b> 4:11 190:8	<b>30</b> 13:17,20 14:4,6	<b>40</b> 79:11 80:3	<b>8.3</b> 297:4 333:17
274:6,10 300:2,6	348:15	14:19,21 25:22	<b>400,000</b> 132:14	<b>8:00</b> 1:9
<b>2(b)</b> 43:20 44:2	<b>24-hour</b> 120:14	36:13 53:2,4,16	<b>44,000</b> 226:3	<b>80</b> 293:2
86:4,8 110:3,6	309:16	54:3,6,9 72:18	<b>47</b> 26:4 143:5	<b>81</b> 76:2
157:1,4 206:3,7	<b>2414</b> 4:13 211:10	75:3 100:3 106:16		<b>82</b> 76:2
240:8,12 279:20	348:17	113:17 130:16	5	<b>85</b> 39:22 241:12
280:2 320:14,17	<b>2496</b> 14:8	218:1 267:6	<b>5</b> 136:12 154:14	<b>87.0</b> 203:2
<b>2,000</b> 132:15	<b>25</b> 89:9 122:19	309:22 311:2	190:10 219:17	<b>88</b> 139:4
<b>2,011</b> 139:2	<b>2502</b> 13:16	343:1 345:5	228:20 252:10	9
<b>2,085</b> 282:11	<b>2503</b> 14:11	349:22 350:6	302:13	<b>90</b> 19:12 74:16
<b>2,327</b> 282:10	<b>2504</b> 14:14	<b>30-case</b> 27:11	<b>5-10</b> 180:20	233:1 241:15
<b>2,400</b> 332:13	<b>2505</b> 14:17 345:4	<b>30-day</b> 4:6,9,16,18	<b>5.1</b> 333:15	<b>90-95</b> 20:5
<b>2.4</b> 135:15	345:13	14:14 18:20 96:5	<b>5.6</b> 191:21	<b>90th</b> 105:5 225:6
<b>2.5</b> 105:8	<b>2510</b> 14:6 346:3	99:3 107:9 128:20	<b>5/95</b> 324:10	<b>94</b> 290:19
<b>2:25</b> 352:14	<b>2512</b> 13:19	131:1 191:20	<b>50</b> 26:7,9 74:18	<b>95</b> 19:12 89:17
<b>2:30</b> 344:8 351:12	<b>2513</b> 4:9 99:3	222:3 264:19	144:4	<b>95</b> 19:12 09:17 <b>95th</b> 324:6
<b>20</b> 35:22 73:7 88:1	128:19 348:19	266:12 288:7	<b>55</b> 26:9	<b>97</b> 40:8 60:5 79:4
88:2 130:19	<b>2514</b> 4:4 57:1,20	345:16 348:10,19	<b>57</b> 4:8	<b>98</b> 40:8
190:13 231:9	58:5 348:13	348:22 349:3,20	6	<b>99</b> 4:10 78:7
233:7 264:21	<b>2515</b> 4:6 57:13 58:5	350:12	<b>6</b> 1:6 4:2 22:19	<b>99.7</b> 203:2
291:16 295:19	58:6 96:4 348:9	<b>30,000</b> 65:2 314:9	128:22 130:17	<b>992</b> 324:1
323:16	<b>2539</b> 4:21 288:18	338:7	135:4 228:20	<b>9th</b> 1:8
<b>20.1</b> 295:22 297:4	326:2 349:7	<b>300</b> 25:8 258:4	258:2 298:16	
<b>200</b> 26:8 27:12	<b>25th</b> 136:6	300-something	200.2 200.10	
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#### <u>CERTIFICATE</u>

This is to certify that the foregoing transcript

In the matter of: All Cause Admissions and Readmissions Steering Committee

Before: NOF

Date: 05-06-2014

Place: Washington, D.C.

was duly recorded and accurately transcribed under my direction; further, that said transcript is a true and accurate record of the proceedings.

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