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1	I	FOOD AND DRUG ADMINISTRATION
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3	FDA-IDSA-NIH-I	Pew Public Workshop
4	Enhancing the	Clinical Trial Enterprise for
5	Antibacterial	Drug Development in the United States
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8	DATE :	Day 2: November 19, 2019
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12		Building 31 Great Room
13		Silver Spring, MD 20993
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		Meeting	November 18, 2019
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1	PROCEEDINGS
2	DAN RUBIN: Good morning, everyone.
3	Please take your seats. I'd like to get started.
4	Welcome to day two of the workshop. We're now
5	beginning our session, strategies to better support
6	antibacterial drug trials. I think we won't go around
7	the table and redo introductions, except I should
8	probably introduce myself since I was in the chairs
9	yesterday and I've jumped all the way to moderating
10	the session along with it. Jan Knisely, who's going
11	to be the co-moderator for today.
12	So I'm Dan Rubin, and I'm a
13	statistician at Cedars. Professor Evans, could you
14	introduce yourself since you weren't here yesterday?
15	SCOTT EVANS: Sure. Good morning,
16	everyone. My name is Scott Evans. I'm a professor
17	and the chair of biostatistics (indiscernible) at
18	George Washington University. And I'm the Director of
19	the SDMC, Statistical Data Metric Center.
20	DAN RUBIN: And do we have any other
21	new people who weren't at the table yesterday? Okay,
22	great. Well, then let's get started. Our first

1	speaker is Dr. John Rex. John Rex is the Chief
2	Medical Officer at F2G''s Manchester, UK based firm
3	and brings 30 plus years of development and policy
4	experienced, focused on antimicrobial agents. His
5	experience puts moving antifungal and antibacterial
6	agents from pre-clinical development through all
7	development phases and various roles.
8	JOHN REX: Thanks, Dan, and thanks to
9	you all of you for being here. And this theme came up
10	yesterday. This is a pivotal moment in time. What do
11	we need, what should we be doing, who should be doing
12	it? And the title of this talk is antibiotic $R\&D$ 3.0,
13	and I'd like us to think about rowing the boat
14	together and in the same direction.
15	These slides are readily available. I
16	actually think the slides from the whole meeting are
17	going to be available, going to be posted, and
18	obviously, I'll share them via my newsletter.
19	So the pivotal point in time is
20	fasting. We have poured enormous resources into pre-
21	clinical and phase one, CARB-X and so forth, and there
22	is really signs of progress. We are seeing some

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Page 6 innovative things deep down in the pre-clinical pipelines, things that nobody had ever discovered They won't all be drugs, but some of them before. will be. Unfortunately, this whole enterprise, hundreds of millions of dollars is going to blow apart due to two intertwined issues: the first is that new antibiotics are kind of like a bridge to nowhere. Actually, that bridge on the left, it actually is, at low tide, there is a path to and from it, but at high tide, it's just out in the middle of the water. And right now, it's high and we're drowning. New antibiotics don't get used. Some of the agents truthfully aren't that interesting. But even the good ones are perceived as only non-inferior. The guidelines are out of date by years and stewardship is based on cost and utility. And then on the flip side is that the payer model is broken. Antibiotics were the fire extinguishers of medicine, as you've heard, and we need to stop paying for them on a per-fire basis. We cannot fix it all at this workshop.

1	The things that are in scope are the types of trials
2	we can conduct, the types of data we can realistically
3	get, how those data should be reported in labeling,
4	how the ID community talks about this data, how the
5	guidelines and the guidance's should handle this data.
6	What's out of scope is payer models. And also, I'm
7	going to say that personally out of scope is, oh, I
8	just don't like it. Okay? If you don't like it, you
9	either have to suggest something else that's better or
10	get with it because you're otherwise keeping us from
11	going anywhere; it's why the boat is spinning in
12	circles.
13	So with that rant over sorry this
14	thing, big picture. Antibiotic R&D 1.0 was sort of
15	the dawn of antibiotics to the mid-2000s. It was
16	generally very easy to see the value in the drugs.
17	But over time, we were learning about weaknesses in
18	our pivotal designs, particularly related to upwards
19	of 20 infections.
20	R&D 2.0 was the moment in time when it
21	all kind of went back to ground zero. The Ketek
22	hearings in Congress in 2007 was a period of time when

	Page 8
1	we didn't know how to develop anything; we weren't
2	sure about it. When we got busy, the f'ing NIH, the
3	whole process led to rapid refinement of not a very
4	orderly science.
5	And we now have very clear, very sound
6	scientific roadmaps for major indications. Single
7	pivotal trials gradually became acceptable for
8	approval, and there was a lot of harmonization between
9	EMA and FDA, and also increasingly recently Japan, and
10	that's all really good stuff.
11	So what's R&D 3.0 going to look like?
12	Well, these are the ideas. It's how do we use LPAD as
13	a springboard, the problems we have to solve, a little
14	bit about superiority again, and the notion that this
15	is a community-level activity and some suggestions.
16	So LPAD, we heard that mentioned
17	yesterday, and Dr. Nambiar talked about three levels
18	of approvals: standard, limited use, and LPAD. In
19	truth, there is no distinct labeling for there's no
20	distinct pathway, it doesn't have a name for limited
21	use; it's just it's the agreement that we can
22	approve with single trials.

	Page 9
1	But LPAD is a special beast. It was
2	created by Congress as part of the 21st Centuries
3	Cares Act, and it really ought to be called LPAAAD
4	because it's limited population of antibacterial and
5	antifungal drug. You'll notice there are least two As
6	missing in that.
7	The concept behind LPAD is that
8	physicians and patients will accept greater
9	uncertainty for serious diseases with unmet needs, a
10	very standard concept, but now codified with respect
11	to antibiotics. And it says that, in brief,
12	streamlined approaches based on severity, rarity and
13	prevalence are what you're going to have to do.
14	And this means single trials, widen
15	inferiority margins, basically being creative within
16	certain boundaries because it's not a license to run
17	riot. You still have to meet the standard of
18	substantial evidence of efficacy based on adequate and
19	well-controlled clinical data.
20	Also, the labeling must make it clear
21	that the limited population that there is a limited
22	population for which this drug is addressed. And this

1	is it's important for us that LPAD gives us two
2	gifts: gift one is the phrase itself, LPAD. This drug
3	is special. Has the logo been designed yet? I don't
4	know, but there's going to be something on the box in
5	a little triangular thing that says this is different;
6	don't use this unless you understand it and you
7	understand the risk/benefit around it.
8	Don't use it in just anybody; use it
9	only in these people. And the package labeling I'm
10	sorry, the label data actually has this phrase right
11	here, limited population in at least six places on
12	page 1 of the label, some of them in big multipoint
13	type.
14	And you combine that with robust
15	stewardship programs and CDC's ongoing surveillance,
16	LPAD agents are something that we can pretty much say
17	to the community, you have to use this right. And we
18	can we should leverage that notion that this is
19	something where we can tell people, be careful here,
20	use it appropriately.
21	But that leaves us with some problems
22	there. R&D 3.0 needs to address two big groups of

	Page 11
1	problems: the first one is communicating the value of
2	standard NI trials, and I put that one separately
3	because that's a community-level thing that we can
4	take on. We have to educate ourselves and our peers
5	on the notion that a cUTI trial is not stupid.
6	And then B, C, and D all tangle
7	together; developing for rare or resistant pathogens,
8	developing for less common infections. And what's the
9	adequate quanta of data for labeling for B and for C -
10	- B, C and D are reduced to study size and how you
11	think about that phrase again, "Substantial evidence
12	of efficacy based on adequate and well-controlled
13	trials."
14	Nowhere in any order or reg does it say
15	that that means alpha means .05 and a margin equals 10
16	percent and a specific end point. None of that is
17	defined. What we do it's up to us to define stuff
18	that is solid enough, and we can consider risk
19	benefit.
20	A little bit of a sidebar about
21	superiority, and you just have to bring this up and

22 remind everybody about this. Superiority is not a

1	generalized answer. If you happen to find superiority
2	for a new drug, great, but that instantly resets the
3	playing field for every subsequent drug. You have to
4	keep that in mind. If it's the core problem is
5	that antibiotic responses are dichotomous cured/not
6	cured. Very few other diseases are that way. I cure
7	your pneumonia. There's nothing better than cured.
8	You go home and have your next 60 years of life.
9	And if it's easy to run a superiority
10	trial, actually something bad has happened in public
11	health resistance must be common enough that I can put
12	people in an arm where there wasn't a good choice.
13	And except for the mildest of infections, superiority
14	means that not only somebody got hurt, somebody
15	probably died and did not really need to die if we had
16	an appropriate drug.
17	So we actually want superiority trials
18	to be impossible. If it's briefly possible due to a
19	gap, the first successful drug closes that gap and
20	makes it hard to go back to doing more superiority.
21	So we all have to know this and we have to teach it to
22	ourselves and teach it to our colleagues: non-

1 inferiority is our main tool; they're sensitive to 2 drug effects. Modern designs work really well, and we 3 have to be very clear about this.

4 And this seques into the notion this is 5 not just a regulatory problem. It's easy to be critical and ask, why is it you say it's nothing; oh, 6 7 it's just a cUTI study. What it was, because you 8 actually want more data. We talked a lot about a lot 9 this, I wish for more stuff. Academia and the journal 10 letters say, well, you did this resistant pathogen 11 study, but it's too small. And the payers say, I 12 expected superiority data in the label. And 13 physicians say, I'll wait for the guidelines to change. And the patients say, non-inferiority sounds 14 15 so dodgy; I don't know what it really means, but it 16 really sounds awful and my doctor doesn't like it. 17 This is all a communication and an 18 education problem as we've just been discussing. 19 People don't understand the core scientific principles. Why is it that some non-inferiority 20 21 trials are rubbish? We have done NI trials and 22 published them in the past and they were rubbish. The

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Page 14 ones we do now are not rubbish. There's a reason for 1 2 that change. 3 Non-traditional agents. We got 4 somebody at the table today who's interested in bacterial phase. There are a variety of other non-5 traditionals we'd like to have developed. They face 6 7 the same problems. We actually -- if we're going to have any of these new toys, we're have to deal with 8 9 those, so here are my suggestions. 10 The first one is: we, as a community, 11 have to become cognizant of the labeling regulations. 12 You need to take the words on this slide and laser 13 etch them on the inside of your eyeglasses so that you're constantly looking at them and are aware of the 14 15 fact that there are things the FDA can do and things 16 the FDA can't do. These are the regulations. An act 17 of Congress would change them, but it doesn't seem 18 terribly likely. Let's work with this. 19 So within -- with that knowledge, supplement as we can. So I would like to think about 20 21 creating working groups to create a credible way to 2.2 work with the available data. We do need to talk

1 about the, in particular, talk about the idea of
2 adequate and the words adequate and well. The patient
3 with the word adequate, remember that patients and
4 physicians will accept different tradeoffs. The word
5 well, I'm not arguing about control, but what's the
6 definition of well.

7 Remember that 100 patients equals \$10 8 million more or less and several years' worth of work. 9 It's not trivial data and it's hard to get all the 10 information you want. Can we supplement with external 11 controls? We're going to hear a talk today about 12 sharing across -- same as yesterday, sharing across 13 body sites. All of this stuff needs to get brought together to talk about what can we do inside the label 14 15 to supplement the label, and at the same time, we need to do this. Agencies, societies and journals need to 16 17 be spreading the words.

18 Superiority trials. I've already 19 talked about that, and just last rant about 20 superiority. Now, you have people who say if it's a 21 superior trial, it would be so much smaller. Yeah, 22 yeah, I know that, right? We all know that. But it's

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1	not a path forward because we try to do those studies,
2	it's lasagnas law. The patients go away because the
3	hospital epidemiologists are making the resistant
4	infection go away. And nobody wants to go to the
5	hospital to test, there's a sign in front that says,
б	world's leading center for resistant bacteria. Come
7	and get your transplant, right? No, no, no, no, you
8	don't want that at all. You want that to not be the
9	case. And this is not a migraine. We're talking
10	about a place where superiority means something bad
11	happened.
12	This is the next message: non-
13	inferiority is not a synonym for worthless, and
14	guidelines need to be or guidances need to be
15	continuously updated. And colistin, please, if I get
16	sick, do not give me that poison; give me a real drug.
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18	Finally, industry needs to be aware of
19	the constraints on the first previous three slides and
20	to focus on novelty and unmet need. You know, Kevin
21	talked yesterday about the need for incentives and
22	different models and that's not today's discussion,

1	but there's a group of us working very hard on pull
2	incentives. But I will tell you this: I think some of
3	those are going to come to be, but they're not going
4	to come to be for every drug.
5	If you're working in this area, it's
6	your job to pick something that you think moves the
7	needle. QIDP alone is not going to move the needle.
8	You've got to actually be doing more than that.
9	So to close at this, you know, at

heart, I'm an ID doc. I moved into industry in 2003 because I spent -- a couple of things: antifungal pipelines were coming to a grinding halt, there was nothing to work on really; and I had spent five years as an epidemiologist, and I was starting to see infections at our hospital that can only be treated with (indiscernible).

I once closed an ICU for about a week because we had an outbreak of a then-untreated infection, and I shut down all the ORs. I said you can't do anything elective, you know, which made a lot of people unhappy. We stomped it out. It was an (indiscernible) bacteria outbreaks. You know, I'm

	Page 18
1	seriously hoping these nasty (indiscernible) will
2	work. But that's the kind of thing that drives us all
3	to be here.
4	Since then, I have walked all sides of
5	this large pharma, small pharma, VC, philanthropic
6	funders, you know, all that. And my deepest message
7	is that tradeoff-free solutions to antimicrobial
8	resistance do not exist; if they did, we'd all use
9	them. Since they don't, we, as a community, it's the
10	time for us to move forward with making the best
11	available tradeoffs, preferably by 1:00 p.m. today.
12	Okay? Thank you.
13	JANE KNISELY: Our next speaker is Dr.
14	Vance Fowler. He's a Professor of Medicine at Duke
15	University School of Medicine. His research interest
16	focuses on staphoreious clinical epidemiology and
17	pathogenesis. And he's led important clinical trials,
18	testing new therapies for staphoreious bacteremia,
19	including a randomized controlled trial comparing
20	daptomycin to standard therapy. And I'm a little
21	surprised that not in his bio, he is also one of the
22	PIs of the Antibacterial Resistance Leadership Group.

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Page 19 VANCE FOWLER: Thanks. These are my disclosures. I'm going to pick up on the point that John made that there are some things that the FDA can do and some things the FDA can't do. And I'm going to pick up from primarily from the perspective of the clinical need. Because the bottom line is, I feel like we in the scientific community are going to have to ultimately generate the data that we need to manage the patients that we see. My points of this talk are outlined here. Registrational trials necessary, not sufficient. Strategy trials, and I'm going to emphasize that hard because I honestly feel like that's giving the people what they want, and in clinical networks to help both approaches. So what do clinicians want? Well, they've told us what they want. This is a paper from 2013 in which a large group of Australian ID physicians were asked what they actually need in clinical trials. And there are some pretty familiar faces here for those of you practicing, per se; joint infection, osteoarticular infection, uncomplicated

staphoreious bacteremia, diabetic foot infection, and
 treating ESDLs and then there were a variation on
 that.

4 And if you actually took these 13 top prioritized items and broke them down into four 5 categories, they were fundamentally listed here: 6 duration of ID antibiotics, how long do we treat these 7 patients; combination drugs, are two drugs better than 8 one; specifically, how do we treat MDR pathogens; and 9 10 then root of administration, can we use PO antibiotics 11 and abbreviate the time a patient has a line in.

These are the four categories, and I'm going to come back to this because this is going to be the benchmark against which I'm going to compare the registrational trials that have been employed and the strategy trials.

So in terms of registrational trials and agents that have been approved from 2013 to '18, and I'll limit my discussion to 2018, so there were the four agents. Again, this is through 2018 for skin and soft tissue, complicated UTI, intrabdominal infection; also acquired pneumonia, the

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Page 21 (indiscernible) from way back in the day, and the (indiscernible). So if you look at the actual trials that were published, phase three registration trials published in 2018, there were basically five. There was the American and Vabor vacuum study, two ABSSSI studies for delafloxacin, and two ABSSSI studies for (indiscernible). So I'll bring you back to that metric of the studies that ID clinicians actually want, none of which were addressed by the registrational trials from 2013 to 2018, so that's my point. And that's not being disrespectful to the FDA; these are just the So now, and the problem brought this up facts. yesterday -- Helen brought it up, I brought it up -you still have to make a decision. So these are three new drugs -- well, cefazoline's not quite so new anymore, but new enough -- taz-avi and (indiscernible). So what I did was I went back for the first quarter of 2018 at Duke, and I asked the question, are these drugs being used, and, if so,

1 where are they being used. And the data shown here is 2 they're about somewhere around 45 or 50 patients, 3 unique patients, who have been treated; 87 percent 4 were being treated off-label, so this is an N1 47 5 times.

And this is the problem right here, 6 7 it's an inconvenient truth, but here we are. Clinical 8 trials don't equal clinical practice. Erythema not receiving in 72 hours alone; to me, that doesn't mean 9 10 failure. Not obtaining a blood culture six weeks 11 after stopping antibiotics for a patient with 12 staphoreious bacteremia, guess what, that's not a 13 failure. And if a patient gets four days of drug B after getting drug A, not a failure. 14

15 So what's the point then? The point is, you know, that we as clinical trialists and we as 16 17 clinicians really are tasked with importantly 18 different responsibilities, both of them essential, 19 but really significantly different. And so, because of that, I'd like to argue that strategy trials really 20 21 go a long way towards addressing the trials clinicians 2.2 want.

1	And I'll do that by demonstrating the
2	strategy trials that were undertaken in 2018, so there
3	were six of them. There was Stephen Harbath's paper
4	on comparing nitrofurantoin for a single dose phosphor
5	for uncomplicated UTI. Colistin monotherapy versus
6	combination therapy for treatment of carbon resistant
7	gram-negatives from Israel. The Danish POET study
8	about using partial oral antibiotics for treatment of
9	endocarditis. The ARREST study which tested the
10	hypothesis that rifampin added to standard therapy
11	improved outcomes in patients with staphoreious
12	bacteremia. The comparison of piperacillin versus
13	meropenem in the Marino trial. And the untesting of
14	an algorithm versus standard care for staphylococcal
15	bacteria. These were all published in 2018.
16	And if you go back to the original grid
17	that I showed about the metrics of what patients
18	what clinicians want and what they're provided, the
19	registrational trials are shown there. All of those
20	categories were addressed by strategy trials published
21	in 2018. So there are differences in these things.
22	We've alluded to it, but sort of to nail it down a bit

1	more refined, the purposes are different.
2	As I see the registrational trial, the
3	primary objective is to make drugs available to
4	practitioners and their patients. Strategy trials are
5	primarily tasked with identifying the best means to
б	use those drugs. The audience is different. At the
7	end of the day, registrational trials, the audience of
8	registrational trials are for regulatory agencies,
9	with good reason.
10	By contrast, strategy trials are
11	primarily tasked with an audience of the scientific
12	community. The study design is different. I talked
13	at length yesterday about the prototypical sort of
14	ABSSSI and the complicated UTI designs and the reason
15	that those well-established protocols are so
16	important; test the drug, not the test.
17	Strategy trials, you're tasked with
18	answering the clinical question. The consequences of
19	failure are dramatically different. If you fail in a
20	registrational trial, the compound is scrapped and the
21	company may close. If you fail in your strategy
22	trial, you publish in class one. The cost is

1	different	and	the	complexity	is	different.	
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So what about networks, how do we fold 2 this in, you know? And one of the challenges I think 3 4 is U.S. site enrollment. Those were really nice presentations yesterday morning with some really clear 5 data on the challenges of U.S. site enrollment. 6 But 7 it's interesting to remember, despite that fact, most 8 clinical patients in clinical trials are enrolled from 9 the U.S. -- this actually came from the FDA website 10 and from a document from 2015 -- and it's not just a 11 little, it's a lot. And that actually is not limited 12 to other diseases, but is specific to infectious 13 disease, as well, so that's point one. Most clinical patients in trials remain enrolling from the U.S. 14 15 The second point, though, is the U.S. site performance in ID clinical trials is variable. 16 17 It really depends on what we're talking about and need 18 to get into the details. So here's an example from 19 complicated UTI -- these are data that were provided to me by a sponsor years ago -- with regards to the 20 21 number of sites, the location of sites and the number 2.2 of patients they enroll. This was for greater than a

1 thousand patients who were enrolled into two phase 2 three identical design complicated UTI trials. Over 3 96 percent of the patients were enrolled outside of 4 the U.S.; out of the 1,060 odd patients, 23 were 5 enrolled from the United States.

Now, let's compare that to another
study. So this is an adjunct therapy for staphoreious
bacteremia that I presented at ECCMID last spring.
Over a hundred patients in the phase two; 80 percent
of those patients were enrolled within the United
States. So then the question is, you know, what makes
a study enrollable in the United States?

13 These are some criteria that I've kind of put together that seems to make sense. First of 14 15 all, the patient's got to be there, and the sort of case in point in that instance is the ABSSSI 16 17 experience. We saw yesterday that ABSSSI was the one 18 indication in which the U.S. continues to increase its 19 numbers, and that's largely due to -- I would argue, largely due to a handful of investigators out in the 20 21 Southwestern United States in the San Diego region. 2.2 And by contrast, in the R pathogens, unless it's ESBR

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or MRSA, I'm sorry, but we don't have enough numbers
in the U.S. to conduct a U.S. limited trial in those
pathogens. And that's a good thing, we want to keep
it that way. It's got to integrate with clinical
practice.

6 The reason these single dose adjuncts 7 were primarily useful or enrollable in the U.S. had to 8 do with the fact that it didn't delay discharge. By contract, a complicated UTI, even in the modified 9 10 quidance now, more or less mandates if it's an IV 11 drug, the patient's got to be in the hospital for five 12 That's not going to happen in 2019 in most days. 13 medical practices, not going to happen in the United They're going out, and you see that in the 14 States. 15 reality that the numbers bear out.

The patient need to own the disease. Somebody alluded to that yesterday in terms of HABP/VABP, and I couldn't agree more. You don't talk to the ID physicians if you want to -- if you're going after a HABP/VABP trial; you talk to folks like (indiscernible). You talk to intensivists because they're the ones who have access to the trial. Now,

1 let's compare that to a complicated UTI. Who owns 2 that? The ambulatory care guy, the ER person, the ID 3 guy, the urologist; where do you get them? You don't, 4 and so we don't. The pathogens need to be there, and 5 we talk about that. ESBL and MRSA, yes, maybe; MDR, 6 not yet.

7 And then I think patient comorbidity 8 needs to be reasonable limit. The VRE example is, I 9 think, a case in point to that from, you know, a 10 decade ago.

11 So some other obstacles to U.S. site-12 based research. And, you know, insufficient trial 13 volume to maintain site infrastructure. You get a 14 cool study. It's, like, wow, this is a really cool 15 study. Where's the coordinator? Oh, you got to go hire a coordinator. Okay. And then I got to assume 16 17 that I'm going to be enroll enough to be able to pay 18 that coordinator just to break even so that the 19 division chief doesn't come back, you know, banging on your head. 20

21 So second point: the site work is 22 academically undervalued. There are a lot of

1	academics in the area, myself included. I know we've
2	all written a lot of letters of promotion. Ask
3	yourself this question: how many times have you ever
4	written in your promotion letter, this person should
5	be promoted to tenure because they were a great site
6	enroller. Don't answer it out loud, but think about
7	it and let me understand. And I think we all probably
8	know the answer: bureaucracy, crushing bureaucracy.
9	And it's both, it's with local and broader.
10	All right. So what are some partial
11	emphasis on partial solutions? I totally agree
12	with John's point that this is a we're in a
13	significant problem here and there's not going to be
14	any single panacea. I think networks can help.
15	Broadly speaking, networks really fall into three
16	flavors: there's an observational type, and I'll use
17	in the case in point, the International Collaboration
18	of Endocarditis. This is one that's very near and
19	dear to my heart but, unfortunately, is not relevant
20	to today's discussion, so this is the closest I'm gong
21	to get to talk about ICE today.
22	Registrational one, and I put the CTTI

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Page 30 initiative up here because I believe this is the first public or published forum in which John Rex and others were able to describe what I think was a really cool idea. And then strategy-based networks; I'm using RLG as an example. The targets are different: observational, describe the disease; registrational, new drugs; strategy, best management. So in terms of the registrational trial, this is, again, this is the paper that John really led, along with Aaron and others, in terms of a vision for how to address the issue of getting new drugs to market. And I'm very pleased that it may have informed somewhat the RFP that Wellcome Trust put out, and I'm really looking forward to hearing more information about that from my colleague. But suffice it to say, the concept would be these are a series of sites focused on a particular disease entity, the sort of three or four that get drugs to market, and that there would be ongoing enrollment with warm-based maintenance such that control patients can be enrolled during that time and you support your infrastructure.

1	The advantages to that is it gives you
2	the right sites for the right trials and that it
3	maintains a trial infrastructure. The disadvantage is
4	critical mass, and then the sponsors may not use it.
5	We're assuming that they will, but at least some of
6	the meetings that I've attended in the past
7	surprisingly, you know, at least gave me reason to
8	suspect that they may not always use these networks.
9	Strategies Trials Network, ARG is one example, there
10	are others. So the purpose of this: design,
11	implement, manage, clinical research in clinical
12	practice.
13	The renewal is I hope will be starting
14	in December. And the three emphasis areas are
15	diagnostics, diagnostics, clinical trials, and
16	relevant science. It's not exclusively focused upon
17	interventional drug studies; and, in fact, I think
18	diagnostics would be an important way to inform the
19	practice as well. There's been you know, we've
20	been productive in the last couple of years, and I'm
21	incredibly proud of the efforts of our collective
22	team.

Page 32
So in summary then, the strategy
registrational trial networks I see fundamental differ
in regards to different functions. Registration, make
new drugs, treatment; strategy, identify new
treatment, best treatment, different audiences,
registrational, FDA/EMA; strategy trials; the
scientific community, different costs and complexity,
so a lot of differences. But there's a similar need
for high quality sites both in the U.S. and abroad,
and I think that's probably an area where alignment
can occur.
So I've sought to make the points that
registrational trials are necessary, non-sufficient
strategy trials, giving the people what they want, and
clinical networks to improve U.S. participation in
both, acknowledge funding. And happy to jump into
questions, and I've been grateful for the time to
speak today. Thank you.
DAN RUBIN: Thank you. Our next
speaker is Pam Tenaerts. She's the Executive Director
at the Clinical Trials Transformation Initiative,
where she works closely with its executive committee

to develop and implement strategies to accomplish
 CTTI's mission.

3 PAM AENAERTS: Good morning, everyone. 4 Happy to be here and thanks for inviting me. So today, I'll be talking about what you can do to 5 enhance enrollment strategies. We heard a lot about 6 enrollment being an issue and clinical trials have 7 8 that specifically, and I'm here to represent the work of the groups that have worked on this. These are my 9 10 disclaimers.

11 And so, just a little bit about CTTI. 12 So we're a public/private partnership, co-founded by 13 Duke University and the Food and Drug Administration, and we have a mission to develop a drug of optional 14 15 practices that will increase quality and efficiency of 16 clinical trials. We basically come up with 17 recommendations on how to do clinical trials better, 18 and that's what we've always done until 2012, and I'll 19 explain a little bit more about that.

Importantly, we involve all the stakeholders, and we're a membership-driven organization. But we involve a lot more organizations

1	than just our members; it's whoever is needed for the
2	activity that we're working on. We're evidence-based
3	in that we use a large social science team to help us
4	go from opinion, because by gosh, when you get a group
5	together on whatever the topic is that they're
6	passionate about, everybody on that group typically
7	knows how to make it better. It's not necessarily
8	evidence-based, so that's what we use our social
9	science team. And we've been lucky enough to have
10	impact in policy documents and things like that.
11	I wanted to highlight a little bit
12	about some of the work we do. We split that up in six
13	areas of focus: quality, talked a little bit about
14	quality by design yesterday about there's resources
15	there; patient engagement is also a big focus of CTTI,
16	which you might have suspected based on my comments
17	yesterday. And then we also help make sure that
18	there's investigators and sites to do the work, even
19	though as you move forward, there'll still always be
20	sites, but that model might be changing a little bit
21	as we go to site lists and things like that.
22	Mobile clinical trials really could

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have fit into a novel clinical trial design, but 1 2 that's just where we use mobile technologies. In our 3 case, it was after consenting, so not to recruit, not 4 to consent, but to capture end points. And then 5 within all the clinical trial designs, you have the antibacterial drug development program; that really is 6 7 the topic of discussion today. We also have a lot of 8 recommendations on the use of single IRB. IRBs came up yesterday too as, you know, the variability in 9 10 approval processes, getting different consents, things 11 like that. 12 So this is our antibacterial drug 13 development. So before this, I would have said CTTI does process improvement on clinical trials, and then, 14 15 oh, and we also do antibacterial drug development. FDA asked us to work on this in 2012, and we developed 16 17 a couple of projects. There's a lot of you in the 18 room here that have helped on these projects, so I 19 want to thank you for that. We could not do this without the people that help us, free most of the 20 21 time, so we really appreciate it.

22

We did three big topics,

1 (indiscernible) HABP/VABP trials, and the reason we
2 picked that one is the reason we're here today, and
3 it's how difficult it is to do these trials. And if
4 we tackle the hardest issues, maybe some of that work
5 could then apply to the other trials as well.
6 Pediatric trials and areas of unmet need.

7 So today, I'm going to talk about the HABP/VABP studies we did, but we also did work on 8 streamlining protocol elements and data collections. 9 10 But the overwhelming thought was it's all great that 11 we can streamline our protocols; that'll help us 12 nothing if we don't have patients in the trial. So 13 that's why we went back to HABP/VABP studies and try to figure out how to do these better, which included 14 15 two portions, a risk factor study and some formative 16 research.

We would really, really, really, really like to do an early enrollment of clinical trial, which is where we test the methodology. And Vance is laughing because we've shopped this around to a lot of people, but we would really like to do this. Haven't been able to do it yet, but it seems really if you

1 want to be evidence-based, you need to have evidence 2 on your methodology, and this would prove it for early 3 enrollment.

4 So why did we talk about early consent, is because there's an early need to treat HABP/VABP; 5 we've learned that today. There are few ongoing or 6 7 planned HABP/VABP. But this must be cyclical because 8 when we started in 2014, there were very few. When we came out with our work on early consent in 2017-'18, 9 10 all of a sudden, there were a couple, and we were, 11 like, oh, we don't want to compete for patients that 12 should go into real treatment trials with our 13 methodology trials, so we put it on hold a little bit. And then now, there's like there's no -- almost no 14 15 HABP/VABP trials again.

So anyway, so what we talked about is and I have that project. And the other thing is, you know, we all know that enrollment rates are, you know, abysmal, they're really low, and the cost of enrolling a patient is really high. We also did work with Tufts University on the cost of trials, patients in HABP/VABP trials and it's about \$100,000 per

1	patient,	this	is	not	trivial,	and	this	is	for	а
2	patient e	enroll	Led	•						

So we looked at that and we had a lot 3 4 of discussions about this. And at one of the meetings, the patient said, well, why don't you ask me 5 to be part of an HABP/VABP study when I'm still there, 6 when I still can consent, and we tossed that around. 7 8 But Vance really turned it into something that was manageable where we could figure out risk factors. 9 10 Because if you want to consent a lot of people up 11 front, you probably don't want to consent a hundred to 12 get one patient that ends up with a pneumonia, but 13 something reasonable that somebody could pay for as far as creating consenting and that. And so, we 14 15 needed to figure out how to get there.

16 So we wanted to do this before they're 17 critically ill because we hope that the patient, the 18 participant could actually participate in the 19 discussion. The family could be there. So by the time the patient down the road has HABP/VABP, is 20 21 potentially on a ventilator, nobody has to sort of 2.2 worry, like, would grandpa really want to do this or

not. You kind of know because you were part of the
 discussion.

And we then figured out how to actually 3 4 do this. So we did ask to do a demonstration study; we haven't been able to do that. So the reason we 5 went to early consent is in our streamlining HABP/VABP 6 7 work, we did a lot of project team discussions and 8 focus groups who'd experienced court leaders, and it really was a challenge to enrolling. The 24-hour 9 10 timeline is the big bugaboo here. And we sort of 11 appealed, like, well, can't we make that 48 hours; for 12 a while we were on the track of let's extend it out to 13 48. And there's been scientific reasons why you shouldn't be doing that, so the 24 really is sort of 14 15 something. But then this early enrollment could really help with that. 16

So even when the patients -- you know, this is prior to effective antibiotic therapy. So if the patient is identified before 24 hours, it's difficult to conduct all these things that you need to do -- consent, labs, study, drug availability -- in that 24 hours. So how can you do this by beginning

consent before the HABP/VABP develops? So you would
 approach and consent patients at high risk before the
 24 hours, before the antibiotics are started, and
 before the symptoms develop.

So you would do all that ahead of time, 5 and then you would enroll the patient the minute they 6 7 would actually develop the diagnosis of -- when you 8 would normally start your planned antibiotics. This is where now, if you had consented early, this is 9 10 where you can now start them in the study instead of 11 starting them on the regular antibiotics that you 12 would normally starting them.

And, like I said, we had planned to conduct a study, which we have not done. If anybody wants to talk about that, I'll be here until 1:00.

So if you want to think about doing this, you have to first find the patients that you would want to consent early, and then figure out whether this is even acceptable or feasible. Because when this plan started coming together, there were a lot of people that said, oh, the IRBs are never going to go for this; we heard that from a lot of people.

1	So we wanted to do both: can we find a
2	set of patients that you could consent that makes
3	economic sense for a sponsor to pay for those
4	consenting and screening procedures; and then, how do
5	you get that done, like, how do you make that
6	acceptable and feasible both for the participants, the
7	science, and the IRBs.
8	So we did some preliminary research;
9	it's really determining the population that you should
10	approach early. And after a lot of planning, we sort
11	of came up with following the oxygen in the ICU,
12	because we figure that running there's plenty of
13	patients that develop pneumonia outside of the ICU,
14	we're very well aware of that. But, I mean,
15	logistically, that's just not an easy thing to start,
16	you know, go after. The ICU is contained, so we
17	figured if we could find a population there, that
18	would make sense for the coordinator to do that.
19	So we identified, we did a risk factor
20	study in the U.S.; the EU participated with a comeback
21	network. And I heard somebody talk about they were
22	participating on our studies as well. And then we

also did pediatric sites because they also have thatissue.

3 So we looked at -- we required invasive 4 or non-invasive ventilation, so this following the oxygen idea, and you had to be receiving antibiotics 5 for suspected pneumonia. So in total, we enrolled 6 7 7500 patients. This is -- I'm going to talk a little 8 bit more about the U.S. one because we're a little further along. Europe took a little longer, and then 9 10 the pediatrics people are working on their section.

11 So this study in the U.S. lasted about 12 eight months in 2016, so this is a pretty short 13 enrollment period. And we identified patients that were high risk, so they had to receive more than 12 14 15 hours of treatment with invasive and non-invasive 16 mechanical ventilation or high levels of oxygen within 17 the past seven days. And of the 7500, 4632 were high 18 risk, identified as high risk; 1400 -- so 1500 19 patients were actually treated for pneumonia; and of those, 539 met the FDA guidance criteria for 20 21 HABP/VABP.

22

So let that sink in a little bit. How

1	many of you have gotten, you know, those cards about
2	my hospital has not had any hospital-acquired
3	infections? So this kind of shows that HABP/VABP is
4	alive and well, so this is another thing to think
5	about. But what we basically found is, you can
6	identify patients that basically if you'd consented
7	10, one of them would have pneumonia.
8	So like I said, we could quasi-predict
9	who might get pneumonia, but you could enrich that
10	sample with other things based on the ICU admission,
11	what had they received. So you can enrich that sample
12	and make it a little richer; documented aspiration
13	risk, admissions source have received of systematic,
14	and the bacterial in the last 90 days were additional
15	risk factors. So if you combine that, you could
16	really identify prospectively patients for an early
17	enrollment strategy.
18	So, okay, that part is done. We can
19	find the patients reasonably that a coordinator could
20	work with to see if they were to develop pneumonia.
21	How about the other thing; is this
22	acceptable and feasible? What concerns would an IRB

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have about this early enrollment strategy? 1 How burdensome would it be to the trial investigators and 2 the study coordinators, because you don't want to make 3 4 it a whole lot harder. And how would patients and careqivers feel about enrolling in a clinical trial, 5 and they don't even have the condition for what's 6 7 going to happen. Not only will they not have the 8 condition, but they're pretty sick already. So here is somebody who is really sick 9 10 coming into an ICU and somebody's going to come to 11 talk to them about, oh, and by the way, not only do 12 you have everything you have, we think you're at high 13 risk for pneumonia. So this is something that people thought may not go over very well. And if we did 14 15 that, what would patients want to know about this approach so they can make an informed decision and not 16 17 sort of a screaming and run away, so we did formative 18 research. 19 So we did in-depth interviews, and then those people were then also asked to participate in 20 21 two surveys. The interviews dealt with the 22 acceptability and preferences for components of that

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strategy and kind of figuring out what topics we
 should describe in the early consent. And then with
 our online survey, we delved deeper into that into
 really getting at the sentences we should be using.
 And then the last survey was final agreement of the
 language to include in the consent.

And the people that participated in this research, in this formative research were patients, caregivers, investigators, study coordinators, and IRBs, so we kind of tried to get everybody that might have an opinion on this.

12 So what the patients and legal 13 representatives thought is that this is really not I mean, they could accept this very 14 that hard. 15 readily; that early consent and enrollment strategy 16 was overwhelmingly accepted. They found it acceptable that their charts would be monitored, because that 17 18 would come with that obviously before the acquired the 19 pneumonia; they could understand the consent information before the -- you know, before they would 20 21 be diagnosed with HABP/VABP, in this case, and they 2.2 would be willing participated under early enrollment

1 trial using approved antibiotics. They felt a little 2 more iffy about the newly ones, but, I mean, we could 3 probably get there, but some others don't work.

4 When we asked investigators and IRB, they kind of felt that this early enrollment strategy, 5 you know, may work. The investigators thought, like, 6 7 oh, this could maybe work. And it may improve 8 efficiency of the trial conduct for HABP/VABP. And we hope to think that this also shows that we might 9 10 potentially be able to do this in other conditions 11 where, you know, there's sort of acute things that 12 happen in a chronic patient.

13 None of the IRB members raised concerns about the early enrollment strategy. So, honestly, 14 15 they said this sounds pretty straightforward, it doesn't sound like it's going to cause a great deal of 16 17 concern. And, you know, there would have to be a 18 discussion about the possibility, the percentages, the 19 chances that might happen, but not looking at this as something unusually concerning. 20

21 What needs to be in that consent ahead 22 of time? So they would want to rationale for the

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early enrollment strategy; they would have to have 1 explanations about non-inferiority. As you can 2 imagine, this was not the easiest concept for anyone 3 4 to understand. And so this was something that you 5 could actually take your time with and close the patient or the perspective participant potentially. 6 7 And their family were there, so this is not like, you 8 know, your family member is on a ventilator and you're freaking out a little bit. 9 10 Reassurances as to what would happen if 11 the studied drug might not be working. We asked them 12 how they would explain that information, and then we 13 used that information to develop and obtain agreement on the text that should be used in the consent, and we 14 15 have finalized texts for all of the issues that they 16 mentioned in the consent. 17 This is available in a publication, and 18 the survey data and the final consent is also going to 19 be published later this year. This got caught up a little bit and, you know, we talked yesterday about 20 21 pre-prints a little bit. And we've starting doing 2.2 pre-print articles because feel that it takes too long

1	to publish, and we are not the hot topic in many
2	cases. You know, we've tried to, you know, go to ID,
3	you know, all these we just find it hard it's
4	hard to publish these things. We finally found a
5	place, but we're going to go more to pre-print. We
6	haven't done it with this project yet, but if we had,
7	it would be available already.
8	We also think that you could use this
9	in other diseases and applications and, you know,
10	other ICU acquired infections or infections that tend
11	to occur, UTIs, (indiscernible), you know, sepsis,
12	other chronic conditions with frequent exasperations
13	like I mentioned earlier, or conditions in which
14	patients have periods of decisional incapacity where
15	for times, they can't make decision maybe so you can
16	talk about this ahead of time.
17	I would say, though, that when we
18	talked to patients and IRBs about this, while they
19	happily wanted to consent ahead of time, they would
20	like a little heads up when it actually would start.
21	Like, it wouldn't be you know, you wouldn't know
22	when you were getting into the research study. So

1 there would be some type of, you know, hey, you
2 remember what we talked about; we're going to start
3 and this is the specific study we're going to put you
4 in.

So this may improve efficiency of 5 clinical trial conduct for HABP/VABP, and it was 6 7 overwhelmingly accepted. Prospectively identifying 8 these patients requires high level of following the oxygen basically, and you could enrich it. And we are 9 10 developing tools to assist with this trial planning; 11 we're going to come out with the consent language. 12 We'll publicly share the risk factor data.

13 We're actually going to put all that clinical data online with levers that if you want to 14 15 look at your inclusion/exclusion criteria, you could 16 sort of move them around on whatever the topic is 17 you're measuring to kind of see that in our 18 population, how that would have affected enrollment, 19 whether you would exclude patients or not. And we're also creating a trial planning -- that's the trial 20 21 planning tool I talked about, and that's it. Thank 22 you.

Page 50 1 JANE KNISELY: Our next speaker is Chibuzor Uchea, a Senior Officer in Wellcome Trust 2 Drug-Resistant Infection Priority Program, working on 3 4 a range of projects focused on the development of new treatments and diagnostics and improving the 5 efficiency of clinical trials. He joined Wellcome in 6 June of 2019 after three years of working in 7 8 healthcare consultancy. 9 CHIBUZOR UCHEA: Good morning, 10 everybody. Thank you for the introduction. I'd like 11 to also thank the organizers for inviting me here to 12 speak. It's a pleasure to be here. 13 So as was mentioned, I'm from the Wellcome Trust. And for those of you who are familiar 14 15 with us, we're a charitable foundation that supports 16 research to improve health around the world. We have 17 a mission of taking on eight global health challenges, 18 making an impact by meeting the response through our 19 priority program. 20 The drug-resistant infections program 21 is one of these, and it aims to use Wellcome's 2.2 funding, convening, advocacy, and influencing power to

1	help lead the global response to antimicrobial
2	resistance. Our program has a strategy that's based
3	on four pillars, which represent key areas of unmet
4	need, but also opportunities in which Wellcome are
5	well placed to be able to make an impact.
6	The four pillars of this strategy can
7	be seen here. I would like to call your attention to
8	the two on the right-hand side, which are most
9	relevant to this workshop: that's the development of
10	new therapeutics, and the acceleration of clinical
11	trials.
12	We have a vision of developing a
12 13	We have a vision of developing a pipeline that's sustainable to develop antibiotics,
13	pipeline that's sustainable to develop antibiotics,
13 14	pipeline that's sustainable to develop antibiotics, diagnostics, and vaccines for infectious diseases to
13 14 15	pipeline that's sustainable to develop antibiotics, diagnostics, and vaccines for infectious diseases to protect local public health. This is reflected by our
13 14 15 16	pipeline that's sustainable to develop antibiotics, diagnostics, and vaccines for infectious diseases to protect local public health. This is reflected by our commitment to CARB-X. As Aaron mentioned yesterday,
13 14 15 16 17	pipeline that's sustainable to develop antibiotics, diagnostics, and vaccines for infectious diseases to protect local public health. This is reflected by our commitment to CARB-X. As Aaron mentioned yesterday, we're a funder and it remains our greatest investment
13 14 15 16 17 18	pipeline that's sustainable to develop antibiotics, diagnostics, and vaccines for infectious diseases to protect local public health. This is reflected by our commitment to CARB-X. As Aaron mentioned yesterday, we're a funder and it remains our greatest investment of \$155 million across five years.
13 14 15 16 17 18 19	pipeline that's sustainable to develop antibiotics, diagnostics, and vaccines for infectious diseases to protect local public health. This is reflected by our commitment to CARB-X. As Aaron mentioned yesterday, we're a funder and it remains our greatest investment of \$155 million across five years. And also our work with the
13 14 15 16 17 18 19 20	<pre>pipeline that's sustainable to develop antibiotics, diagnostics, and vaccines for infectious diseases to protect local public health. This is reflected by our commitment to CARB-X. As Aaron mentioned yesterday, we're a funder and it remains our greatest investment of \$155 million across five years. And also our work with the (indiscernible) program in collaboration with the</pre>

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1 tighter bottleneck, and we're committed to developing initiatives that will help accelerate (indiscernible). 2 We also have a strong policy and 3 4 advocacy focus, and we interact with key decision makers in all of our activities as we recognize that 5 antimicrobial resistance has an urgent need for global 6 7 action. 8 So the current funding model of clinical development in infectious diseases is complex 9 10 and burdened with a range of inefficiencies that slow down the commercialization of antibiotics. This is 11 12 especially true in low- and middle-income countries, 13 and is reflected by the disparity between the number of trials that were run in these countries and the 14 15 burden of antimicrobial resistance. 16 A key driver of this is the ad hoc funding nature of individual trials, whereas sponsors 17 18 custom build a single use network of trial sites, 19 spending considerable amounts of money building 20 capacity and infrastructure and training staff on 21 individual protocols. At the end of these trials, the 2.2 capacity is disbanded, and there's a loss of

infrastructure and expertise within sites as well.
 Subsequent funders then come in and spend considerable
 amounts of money reactivating sites, redeveloping
 infrastructure, and training new members of staff on
 subsequent protocols.

Another key issue, as mentioned 6 7 yesterday, are issues with patients recruitment. And 8 because of the clinical nature of infectious diseases, there's a very narrow window for recruitment. 9 And 10 also, as described yesterday, it's not only impossible 11 to be able to move patients to various different 12 sites.

13 Also, if the focus on studies in high income countries, there's difficulty in finding 14 15 patients with -- some of them more difficult to treat infections, especially if they're multi-drug resistant 16 17 and extensively drug-resistant indications. Whereas, 18 there were a much wider pool of patients in low- and middle-income countries with these extensive drug-19 20 resistant infections because of the increased burden of antimicrobial resistance. 21

22

And in low- and middle-income

1	countries, there were also issues with the
2	(indiscernible) of trial site quality, which affects
3	the quality of the data that it produced, and also the
4	requirement for individual studies for each indication
5	exacerbates the issues with our funding model.
6	So what are the proposed solutions for
7	these problems? We have a vision of accelerating
8	clinical development in this space using a two-pronged
9	approach that will address the various different
10	barriers.
11	The first of these is the development
12	of international clinical trial networks, which will
13	strengthen the clinical trial capabilities within low-
14	and middle-income countries, alleviating the
15	inefficiencies of the current ad hoc funding model and
16	vitally provide access to large populations of
17	patients with key drug-resistant infections.
18	We have a vision of these networks
19	being able to run multiple studies from different
20	funders simultaneously.
21	Our second approach is the development
22	of platforms for innovative trial design, especially

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1	the use of continuous master protocols. This will
2	allow control group sharing between trials and reduce
3	costs and burdens.
4	These two aspects are independent, but
5	also complementary, and joint implementation will
6	provide even greater efficiencies.
7	So the first of these is the
8	international clinical trial networks. And we're
9	working towards developing a pilot network that's
10	going to be entered in Southeast Asia. It will start
11	as a flexible scalable regional network of high-
12	quality sites, which it will be building on existing
13	capacity, and vitally will provide rapid access to the
14	Southeast Asian population with a high burden of
15	antimicrobial resistance.
16	Our model will be (indiscernible) with
17	other regional networks, which will be vital for us to
18	be able to broaden our base, and also will provide
19	access to key expertise in infectious diseases
20	internationally. Another key aspect of this is that
21	we would be able to leverage key learnings from the
22	development of other clinical trial networks.

	rage 50
1	Our business model will be orientated
2	to be able to engage with the private sector, which
3	will allow us the support by investigating initiated
4	studies and registrational studies. We've identified
5	our nucleus of sites who will be our founding members,
6	and we aim to launch the network around an initial
7	trial, which will be used to test and help develop the
8	network and also inform scale.
9	An effective clinical trial network
10	will provide a range of benefits to a number of
11	stakeholders, including sponsors, investigators, and
12	ultimately patients. For sponsors, there will be
13	reduced costs of conducting trials through warm-based
14	benefits, and also will facilitate power around follow
15	on studies and optimization studies for strategy
16	trials as (indiscernible)'s ability.
17	For investigators, crucially will
18	provide access to keep populations with high burden of
19	disease, and also will improve the quality of trial
20	sites in the region, improving the data that are
21	produced and helping to produce better studies.
22	And ultimately for patients, it will

increase the speed at which treatments have become
 available, and also potentially make them cheaper
 through reduced developmental costs.

We've recently commissioned a piece of work, which is looking at how best to structure the network to maximize our possibilities of success. And the process we engaged in with a range of stakeholders from industry, clinical trial experts, as well as other members from other clinical trial networks.

10 Our own grading model will be with the 11 network secretariat, which is responsible for managing 12 and coordinating central functions, and also it will 13 be guided by a steering committee which will provide strategic oversight. The responsibilities and roles 14 15 of the secretariat include pipeline management and regulatory engagement and market access, which will be 16 vital for the long-term sustainability of the network, 17 18 as well as administrative functions.

We've also identified key areas which are vital for startup, and also others which are more important for the long-term vision of the network. These include geography; we've identified our founding

1	members, but we are also looking to expand, and we can
2	add additional sites that match the site quality
3	criteria to the network. We're currently engaged in
4	conversations with sites and institutions in India,
5	which will further increase our geography and vitally
6	access to either ratable patients.
7	The network will start by looking at,
8	investigate and initiate treatment trials within drug-
9	resistant infections, but we'll scale up to include
10	studies from sponsors in a commercial sector and also
11	looking at infectious diseases in general.
12	We're currently working to finalize our
12 13	We're currently working to finalize our governance structure, which will be ready at the end
13	governance structure, which will be ready at the end
13 14	governance structure, which will be ready at the end of the year, and then we will start to recruit for our
13 14 15	governance structure, which will be ready at the end of the year, and then we will start to recruit for our secretariat function, as well as identifying and
13 14 15 16	governance structure, which will be ready at the end of the year, and then we will start to recruit for our secretariat function, as well as identifying and developing the initial clinical trial, which will be
13 14 15 16 17	governance structure, which will be ready at the end of the year, and then we will start to recruit for our secretariat function, as well as identifying and developing the initial clinical trial, which will be across next year.
13 14 15 16 17 18	governance structure, which will be ready at the end of the year, and then we will start to recruit for our secretariat function, as well as identifying and developing the initial clinical trial, which will be across next year. There are a number of potential
13 14 15 16 17 18 19	governance structure, which will be ready at the end of the year, and then we will start to recruit for our secretariat function, as well as identifying and developing the initial clinical trial, which will be across next year. There are a number of potential challenges of developing international clinical trial
13 14 15 16 17 18 19 20	<pre>governance structure, which will be ready at the end of the year, and then we will start to recruit for our secretariat function, as well as identifying and developing the initial clinical trial, which will be across next year. There are a number of potential challenges of developing international clinical trial networks, and these include a strong and robust and</pre>

1	Another key aspect is potential for
2	underuse of the network between studies. This is why
3	to prevent a lot of times sustainability of the
4	network. And the work we've commissioned has looked
5	at the pipeline. And although there are considerable
6	studies that can come into the network, there's also
7	the opportunity in which there will be down time
8	during when registrational studies are not
9	available. For this reason, we will work for our
10	network to be able to have the capacity and the
11	expertise to run additional studies, especially more
12	complicated studies, including pediatric studies and
13	smaller optimization studies.
14	We're currently working to build out
15	our secretariat function, which will provide support
16	for the network to be able to address these
17	challenges.
18	So the second approach is the use of
19	innovative trial design. In 2016, we held a workshop
20	assessing the benefits or potential benefits of using
21	clinical trial networks for antibiotic development,
22	which a number of you would have attended. An

1 innovative trial design through the use of continuous 2 master protocols was highlighted as a key tool that 3 could further accelerate the efficiencies that are 4 provided through clinical trials, and a recommendation 5 was made that this should be further explored.

Since then, we've commissioned some 6 work to assess the feasibility of using continuous 7 8 master protocols within clinical trial networks. And 9 in the process, we engaged with key stakeholders in 10 the industry, regulatory bodies, and John Rex has also 11 been a great guide in our thinking. The continuous 12 master protocol will provide even greater efficiencies 13 by allowing the sharing of control groups within clinical trials with the same indications, which will 14 15 reduce the enrollment periods and reduce the costs as 16 well.

There's the possibility to use concurrent control groups within the same time periods have also potentially non-concurrent controls and possibly control data, historic control data from previous trials. And the use of adaptive randomization allows the flexibility of entering and

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1	exit of compounds throughout the process, mitigating
2	the irregular startup of regulatory studies.
3	Our feasibility assessment has shown
4	that such a platform would be especially beneficial
5	for undertaking regulatory science studies and later,
6	expansion studies. We have identified sufficient
7	demand for this among the stakeholders which we've
8	engaged with.
9	The use of concurrent controls has been
10	supported in principal by regulators. However, there
11	remains considerable concerns and issues around the
12	use of long concurrent controls as we discussed
13	yesterday.
14	We have identified adult HAP/VAP and
15	CAP studies, as well as pediatric studies as those
16	that will benefit most from the use of the continuous
17	master protocol, and we're currently exploring with
18	other funders whether this type of platform is
19	investible.
20	So in conclusion, we believe that the
21	reason strengthening of the pipeline really demands a
22	more efficient clinical development process. And the

initiatives that we've mentioned here today around the 1 use of clinical trial networks and continuous master 2 protocols will provide initiatives -- will provide 3 4 additional benefits from a scientific, financial, and 5 developmental perspective. Importantly, conducing these in low-6 and middle-income countries will provide access to key 7 8 populations of patients with drug-resistant infections. 9 10 Thank you very much for the DAN RUBIN: 11 introduction. I'll try to keep my remarks brief to 12 leave time for other speakers and for Q&A and to keep 13 us on schedule. One thing I might say is that after 14 15 this stat session, we have some Q&A scheduled for all those talks. We didn't have any on the agenda for Q&A 16 17 for the talks we just heard this morning. But maybe 18 after all the stat speakers, we can have a Q&A for everything so far today and then go into the panel 19 discussion later this afternoon. 20 21 What I'd like to do is provide some

high-level comments about three important statistical

2.2

Page 63 issues in anti-infective registration trials, which I 1 hope will frame the discussion for this session. 2 These issues are endpoints, borrowing information 3 4 across body sites, and considerations for carbapenemresistant pathogen studies. 5 With respect to endpoints, anti-6 7 infective registration trials have most commonly used 8 binary endpoints in which each outcome is classified as a success or failure. The definitions recommended 9 10 in our guidance document vary across the disease 11 types. And hospital-acquired and ventilator-12 associated bacterial pneumonia, the main endpoint 13 recommended in the guidance is all-cause mortality at a fixed time point following randomization between day 14 14 and day 28. In trials of intra-abdominal 15 infections, failure is based on a clinical 16 determination of failure. And in complicated urinary 17 18 tract infections, the primary endpoint is a composite 19 based on both clinical and microbiological components. 20 Professor Evans will describe in more 21 detail how binary endpoints could potentially be 2.2 improved by moving to endpoints with finer gradations

	rage or
1	of success and failure. For instance, one could
2	consider an ordinal endpoint with the three levels of
3	death, survival with major morbidity, and survival
4	without major morbidity. The main advantage is that
5	this may lead to more informative comparisons when
6	reasons for failure are not all lumped together. And
7	another advantage of this approach is that it can
8	increase statistical power because it's making use of
9	more information.
10	One consideration is that levels should
11	be chosen so that differences will not be solely
12	driven by effects on components with minor importance
13	or solely by safety.
14	As a hypothetical example, if you
15	wanted to show that a new drug was superior to say
16	colistin and included reversible renal toxicity in an
17	ordinal outcome, one thing to check would be whether a
18	new drug could be superior to colistin using this type
19	of outcome scale even if it was say increasing
20	mortality. And this issue can be addressed to some
21	extent by assigning weights or utilities to different
22	categories.

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1	My understanding is that these forms of
2	DOOR and RADAR methods have been used by ARLG, have
3	mainly been proposed and implemented for superiority
4	trials.
5	Dr. Lewis will discuss borrowing
6	information across body sites of infection. Some
7	statistical methods such as Bayesian hierarchical
8	models have been widely used for combining information
9	from different sources, both in clinical trials and
10	many other types of applications, but largely not in
11	the anti-infective space.
12	The models would attempt to provide an
13	integrated or synthesized analysis of patients with
14	different infection types.
15	One way to think about information
16	borrowing is that it tries to reduce noise for
17	estimating treatment effects in any one body site by
18	bringing in other relevant data. So for instance,
19	suppose you have a very small number of patients with
20	complicated intra-abdominal infections so that the
21	estimated treatment effect will have a lot of
22	variability. If there's a larger number of patients

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with complicated urinary tract infections who have
been treated with the same drug, then incorporating
them into the analysis could in theory reduce this
variability.
Of course the issue with doing this is
that there is a reliance on modeling assumptions about
the degree of similarity between infection types. So
for estimating treatment effects in cIAI, you would no
longer have the protection of a fully unbiased,
randomized comparison.
One remark to point out is that FDA has
at least informally used some forms of information
borrowing in assessing anti-infective drugs even
without use the full Bayesian machinery. For
instance, an NDA can be based on a single successful
Phase 3 trial for patients at one body site with
supportive evidence from a related disease.
Registration trials do combing
heterogenous patients, such as studies of both
hospital-acquired and ventilator-associated bacterial
pneumonia. And our unmet need guidance also accepts
pooling body sites for superiority trials, and

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1	registration trials in resistant pathogen studies have
2	combined nosocomial pneumonia and bloodstream
3	infections.
4	There are a few issues that FDA is
5	likely to take into account when reviewing proposals
6	for an integrated analysis of body sites. One issue
7	would be statistical operating characteristics if
8	treatment effects differ.
9	For instance, if there is a proposal to
10	combine HABP/VABP, cIAI, and cUTI, and in truth the
11	drug doesn't work in HABP/VABP, how inflated is the
12	chance that the design will lead to a false conclusion
13	of efficacy in HABP/VABP?
14	We may also consider previous history
15	of discordant results of cross-body sites, such as
16	with daptomycin not working well in pneumonia due to
17	inactivation by pulmonary surfactant. And for
18	deciding which data to integrate, this wouldn't just
19	be a statistical decision; there would of course also
20	need to be a clinical judgement that the infection
21	types, pathogens, and endpoints make sense to combine
22	in an integrated analysis.

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1	The last topic I'll touch upon are
2	carbapenem-resistant pathogen studies, which are
3	related to Dr. Dane's upcoming presentation. These
4	studies are commendable because they do directly
5	address questions most closely related to unmet
6	medical needs.
7	The main challenge is that enrollment
8	in randomized trials has been very difficult, as you
9	have heard over the course of the workshop, due in
10	part to the rarity of the pathogens.
11	I will just briefly note that there are
12	some government or academic randomized trials that
13	have compared combination therapy to colistin
14	monotherapy in patients with carbapenem-resistant
15	Acinetobacter baumanni infections, which have had some
16	degree of success with enrollment and together have
17	enrolled about 650 patients. And that's not including
18	the OVERCOME study we heard about yesterday from Dr.
19	Dixon. But I've referenced a metanalysis on this
20	second bullet.
21	If moving beyond randomized comparisons
22	to external controls or non-randomized comparisons,

the key challenge is being able to control for
 confounding in patient populations with many comorbid
 conditions. Dr. Dan will go into more details about
 some of these considerations surrounding randomized
 and non-randomized evidence.

Our unmet need guidance also discusses 6 7 pathways that have now been used by several sponsors 8 based on another type of information borrowing. And this is where a non-inferiority trial is conducted in 9 10 patients with susceptible pathogens, potentially with 11 a wider than normal noninferiority margin. The 12 labeling is then for patients with more limited 13 treatment options with the constraints of the data package communicated in the labeling. 14 15 The main uncertainty of this approach 16 relates to differences between patients with susceptible pathogens and the less well-characterized 17

18 group with limited treatment options.

19 One question to consider about this 20 form of extrapolation is whether the discordant 21 directionality of numerical results in the carbapenem-22 resistant CARE trial of plazomicin and the CREDIBLE-CR

study of cefiderocol could have been predicted or
 explained based on all the preclinical information or
 from the successful non-inferiority trials conducted
 for each drug in carbapenem susceptible complicated
 UTI.

6 The caveat is that the numerical 7 results in these two trials of carbapenem-resistant 8 infections had statistical noise from the sample 9 sizes. But the references in this bullet are to our 10 advisory committee materials if you'd like to think 11 more about this question.

12 The last point I'll make about 13 carbapenem-resistant pathogen studies is that folding 14 these patients into standard noninferiority trials but 15 with more flexibility in the active comparator would 16 follow the template used for previous types of 17 resistance.

For instance, if evaluating a grampositive drug, there wouldn't be a separate trial for MRSA. Patients with MRSA instead would most likely simply be included in a noninferiority trial of skin infections or pneumonia or bacteremia and there would

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be provisions for patients in the control group to
receive an agent with activity against MRSA.
The uncertainty or one of the main
uncertainties that I see with using this approach for
carbapenem-resistant infections is that any recently-
approved active comparator might itself have been
studied using relatively streamlined data. And this
could limit the interpretability of noninferiority
conclusions if there were remaining unanswered
questions about the active control. And one term that
has been used for this type of risk is biocreep.
Here are references for several of the
topics that I've mentioned, and thank you.
JANE KNISELY: Thanks. Our next
speaker is Dr. Roger Lewis. He is a Professor and
Chair of the Department of Emergency Medicine at
Harbor-UCLA and a Senior Medical Scientist at Berry
Consultants. And he will be speaking about Bayesian
adaptive clinical trials.
ROGER LEWIS: Great. Thank you very
much. It's a pleasure to be here. Thank you to the
organizers and the audience.

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1	So I'm going to be talking about
2	statistical approaches for antibiotic trials and
3	posing the question, are we answering the wrong
4	questions? But in posing that question, I hope to
5	follow Dr. Rex's recommendation to actually give an
6	alternative.
7	These are my disclosures.
8	So the take-home points are that the
9	statistical and overall design strategies that are
10	commonly used for antibiotic trials really should
11	directly inform the clinical uses of products if they
12	are approved. In light of how antimicrobials are
13	actually used, I believe they should support
14	antibiotic stewardship to the degree available or
15	accessible within the regulations, and also inform
16	regulatory approvals and labeling.
17	I put these points in this order
18	because I believe that if you have trials that
19	directly address the appropriate clinical use of an
20	agent, that will naturally inform regulatory decision-
21	making. But I believe as currently deigned, many
22	trials in antibiotics do not meet these goals. They

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1	are overly-narrowly focused. We heard amazing ratios
2	of screen-to-enrollment numbers yesterday. They may
3	predictably undermine some stewardship efforts, and
4	they risk missing important benefits due to delayed
5	initiation of agents and contamination of patients
6	'outcomes by prior treatment, but they could address
7	the goals.
8	So let's briefly go through an anatomy
9	of an antibiotic trial as sometimes carried out. And
10	I hope that I don't step on anybody's toes this way.
11	The patient presents commonly to an
12	emergency department, which is where I work
13	clinically. They have signs and symptoms of a serious
14	bacterial infection. They are hopefully briefly
15	evaluated, and empiric antibiotic therapy is started
16	very early because there is an important clinical and
17	compliance imperative to do so. There is a small loss
18	of patients if it's determined that treating the
19	disease is not in the patient's interest and
20	consistent with goals of care.
21	Some time after that, they are admitted
22	to the hospital. And it's commonly in the in-hospital

1	setting in which they are first evaluated for an
2	antibiotic trial. At some point it's confirmed that
3	the intended single site of infection for the trial is
4	in fact the infected site. Culture results come back
5	that meet the criteria. At some point they're
6	actually randomized. And if they are successfully
7	enrolled and randomized, the investigational medical
8	product is begun, and at some point there's an outcome
9	assessment.
10	There's a long period of time where
11	they are receiving empiric therapy prior to
12	randomization, and lots of things happen during that
13	time that may dilute the true treatment benefit of the
14	investigational product. So the limitations of this
15	approach include little alignment with clinical
16	practice, a narrowly-defined study population,
17	relative late initiation of the investigational agent,
18	and the evidence that is gathered and therefore the
19	labeling that can result from the trial addresses a
20	single infection site. So how can we address some of
21	these challenges?
22	Another way of pointing out the

1 differences between the trial structure and clinical 2 care is to look at the timing of the treatment, the populations that are included, the motivating event 3 4 that leads to the use of either the investigational product or the approved product in clinical use, the 5 types of infections that are treated, and whether 6 we're looking at non-inferiority or superiority. All 7 8 of those things are relatively different between the clinical use and the way we study these agents. 9 And 10 those can increase the risk that we produce data that 11 doesn't allow us to make the best-possible regulatory 12 decisions. 13 So clearly new agents are needed to

14 treat challenging organisms across multiple sites of 15 infection. And in clinical practice, antibiotics with 16 demonstrated penetration into the infected site and an 17 appropriate antibacterial coverage are routinely used 18 for infections at those sites, independent of any 19 site-specific supporting data, and certainly 20 independent of labeling.

Now, there has been, as we all know,
problems with a surprising lack of antibiotic efficacy

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at specific sites, most notably lung. And that's
generated concern regarding sharing efficacy data
across anatomic sites. When I talked to people who
are deeply embedded in this area, there is almost an
air of PTSD regarding the trials that resulted in
these surprising findings. And that experience should
not cause us to abandon the common clinical reasoning
that an antibiotic that works well in multiple sites
is more likely to work well at another site than an
antibiotic that does not work well at multiple sites,
but we also need to do so with our eyes wide open.
So the proposed strategy I'm going to
discuss is a platform trial with enrollment timing and
antibiotic initiation designed to match clinical use,
careful integration of information across body sites -
- and we'll spend quite a bit of time talking about
what careful means that simultaneously addresses
both noninferiority and superiority, because I think
that choice is a false dichotomy, and may achieve
additional efficiencies through the platform trial
structure.
So the proposed trial would look

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1	something like this in terms of flow. The patient
2	presents with the signs and symptoms of a serious
3	bacterial infection and is at risk for but does not
4	have a documented infection with a highly-resistant
5	organism. The patient is evaluated for participation
6	in the trial as part of their emergency department
7	evaluation, and the first antibiotic they receive is
8	the randomized investigational product. At some point
9	after that, they're admitted. And at some point later
10	we get the results of the cultures, additional
11	diagnostic tests that verify the actual site of
12	infection and the resistance pattern of the organism
13	or the fact that there is no isolate identified, and
14	then the outcome is assessed.
15	And the question is how could we use
16	data from this type of simpler structure to
17	appropriately inform both regulatory decision-making
18	and clinical decision-making regarding use of approved
19	products. The advantages is that this aligns with
20	clinical practice, there's a broad study population,
21	there's early initiation of the investigational agent,
22	and you will get information that informs you about

evidence across resistance patterns and across
 clinical sites of infection.

So first, what do we mean when we talk 3 4 about a platform trial? So a platform trial is an experimental infrastructure that's intended to 5 evaluate multiple treatments, often for a group of 6 7 diseases. For a group of diseases here, you can think 8 of different infection sites or different underlying pathogens. And the platform trial is intended to 9 10 continue beyond the evaluation of any individual 11 treatment. The treatments are often used in 12 The trial explicitly incorporates the combinations. 13 idea that the diseases that are being studied are similar to each other but not identical, and there's a 14 15 dynamic list of treatments that are available. 16 So this terminology is borrowed from

17 the oncology world. And I first want to make a 18 distinction between a master protocol and a platform 19 trial. The term master protocol is a very broad term 20 that may include simply the use of very standardized 21 clinical processes to increase the efficiency of a 22 clinical trial. But the separate experiments within

Page 79 1 the master protocol may be inferentially separate. You may be asking separate questions about completely 2 separable populations or treatments. 3 4 A platform trial in contrast in my view incorporates some sort of statistical sharing of 5 information to increase the efficiency of the 6 7 inference given any particular set of data. An umbrella trial is a term used then 8 you're testing multiple different drugs for diseases 9 10 that occur at the same site. And you may think of 11 that as multiple studies of a pneumonia with different 12 underlying etiologies or organisms. 13 A basket trial is a trial in which the disease shares a commonality that is a target of the 14 15 treatment, but is in different anatomic sites. That would be for example sharing of information across 16 17 body sites when you're using a single drug that's 18 active against a particular class of organism. 19 There's a very nice review article 20 written by Drs. Woodcoock and LaVange that goes 21 through this terminology and provides a very nice 2.2 background.

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1	So what is the proposed strategy? The
2	patient presents with an infection in either the lung,
3	the urinary tract, or the abdomen. But if you're
4	interested in gram-positives, you could replace this
5	figure with the places where gram-positives are of
6	most interest and challenging. And from each of these
7	sites, we obtain bacterial isolates, and we either
8	identify them as being sensitive to the standard of
9	care, resistant, or unable to be cultured. We want to
10	demonstrate superiority to the standard of care among
11	those patients who have resistant isolates. Non-
12	inferiority to the standard of care control arm in
13	those that are sensitive to the standard of care. And
14	I put a question in there what you would do with those
15	who have missing bacterial isolates. And we want to
16	take advantage of the multiple body sites.
17	Over time, the agents that are being
18	considered could also change. One could start with a
19	control arm and maybe one, two, or three active arms.
20	And over time, additional arms can become available.
21	You may decide that based on additional safety data or
22	other preclinical information, it becomes appropriate

1	to combine arms into a single arm. That's the A plus
2	D. Arms may be discontinued for harm or futility at
3	any point. And if one of the arms is demonstrated to
4	be superior, that can seamlessly become the new
5	standard of care in a platform trial without
6	redesigning the trial. This addresses a point that
7	John made about what happens if you identify a new
8	standard of care; do you suddenly undermine your
9	ongoing research efforts?
10	This approach has been modeled
11	numerically through work that was supposed by the ARLG
12	and others and involved a wide variety of
13	collaborators. And in general if you consider this
14	approach and you make reasonable but particular
15	assumptions about the number of drugs, the fraction
16	that are positive or negative, et cetera, you can save
17	an average of something between 40 and 60 percent in
18	sample size per answer you get regarding the
19	superiority of the drug and resistance, isolates, or
20	non-inferiority in sensitive isolates. And this
21	efficiency comes from the shared control, the sharing
22	of information across body sites, and the fact that

1 the study can seamlessly drop an arm for inferiority 2 or for futility. And the assumption is that another 3 drug is available to be tested. So that's a very 4 particular assumption.

I want to specifically comment about 5 the sharing of information between body sites. 6 So we 7 all know that antimicrobial agents are likely to have 8 different effects at different body sites, for a wide variety of reasons. The question is whether the 9 10 degree of variability in the treatment effect is 11 clinically unimportant and we should think about the 12 treatment effects as similar across body sites, or 13 it's clinically important so we should think differently about the use of that drug across body 14 15 sites.

So the question is how can we address both the possibilities that the treatment effects will be largely similar and that the treatment effects will be highly disparate in a single statistical model.

20 Clinicals do this sort of borrowing all 21 the time. We take data that have been obtained in 22 particular types of patients and we apply them to

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1	others. But we do this in a way that is not well-
2	documented and is not quantitative. Here we need a
3	strategy that is completely prespecified, that is
4	statistically rigorous, and whose operating
5	characteristics can be evaluated numerically.
6	It is very tempting to try to avoid the
7	reality that similarity is not an all-or-none thing.
8	We tend to behave as if it is. So we will group all
9	types of infections together if they were included
10	within the inclusion criteria for a trial, or we will
11	exclude them if they were not. We will include, for
12	example, all different sorts of isolates if they were
13	at least represented at all within the trail.
14	Although we certainly don't have enough evidence to
15	make independent estimates of the efficacy of the
16	agent across each species of bacterium. For example,
17	it was isolated.
18	The all-or-none approach puts us at
19	tremendous risk of failing to identify subgroups that
20	do experience different treatment effects of
21	complications if we start by combining them all
22	together. So we miss real differences that exist.

And we also fail when we separate them to recognize compelling circumstantial evidence of treatment efficacy. For example, if the drug works across four or five different infection sites, the chance it will work against the fifth or the sixth infection site is greatly increased.

8 is a Bayesian hierarchical model. And I've underlined the key conceptual strength of this approach. 9 The 10 Bayesian hierarchical model shares information across 11 subgroups, or in this case, body types, to the degree 12 that is justified by the consistency of the 13 information. So, contrary to something that is commonly stated, that you have to make an assumption 14 15 about how similar the treatment effect is across body 16 sites, instead what you do is you create a model that 17 can learn how consistent the treatment effect is 18 across body sites and use a greater degree of pooling 19 if you see consistency and a lesser degree of pooling if you see heterogeneity of the treatment effect. 20 So 21 how does that work?

22

So consider a data set in which you're

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enrolling three subtypes of infection and you get 1 three separate point estimates if you simply looked at 2 the treatment efficacy that appears to be present in 3 4 each one of those three sites. Each one of those point estimates has some uncertainty around it. And 5 the hierarchical model assumes that those treatment 6 7 effects are themselves drawn from a population of 8 treatment effects among lots of different body sites that might exist. And the hierarchy is an assumed 9 10 distribution of body sites or subgroups, and there's 11 uncertainty in how similar those subgroups will be to 12 That's the prior variability at the third each other. 13 level of the hierarchy. So you actually allow the model to realize you don't know how similar the 14 15 treatment effect will be in the three body sites or the four body sites. 16 17 When the hierarchical model fits the 18 data -- and this is sometimes called a shrinkage 19 estimate -- it creates new estimates for the treatment effect in each of the body sites individually that 20

21 learns from the consistency of the effect across body 22 sites.

Page 86 1 There is a very general statistical fact called the James Stein effect -- I put the 1961 2 publication there just to show it's not new -- that 3 4 says that the best estimate of a true treatment effect 5 in a subgroup is actually not the treatment effect you get by simply looking at the data within the subgroup 6 7 if there are three or more subgroups. This is a very 8 general, non-Bayesian result that is horribly, horribly inconvenient because it is so 9 10 counterintuitive. It means that when you look at the 11 treatment effect in a type of disease for a drug, you 12 should think about how well that drug works in all 13 kinds of other similar diseases, just like clinicians do every day. 14 15 So how about the non-inferiority and superiority issue? The data that one would get from 16 17 this proposed approach to a platform trial would 18 include data when the infecting organism is isolated 19 and found to be highly resistant, when it's found not to be highly resistant, and also you're going to have 20 21 some patients who clinically appear infected for which 2.2 you are unable to identify a specific etiologic

1	isolate.
2	The first data set from the patients
3	who have isolates that are highly resistant can be
4	used primarily to address a superiority hypothesis.
5	It can secondarily be used to address a non-
6	inferiority hypothesis if there is value in approving
7	a drug simply because it's non-inferior to the
8	comparator in that setting.
9	When the infecting organism is isolated
10	but found not to be highly resistant, that can be used
11	to address a non-inferiority hypothesis and also to
12	address safety and PK and other goals.
13	When the patient has no isolate
14	obtained, that can be used to address the non-
15	inferiority hypothesis, but there are some very
16	specific considerations there depending on the site
17	and what a non a positive culture means in that
18	setting. That would be a great thing to discuss at
19	length another day.
20	But the point here is that the
21	membership of the patient into each of these subgroups
22	is based on a pre-treatment assessment, which is the

1 culture. So these are valid subgroups for which you 2 can draw valid statistical inferences. This approach 3 allows the early initiation of the investigational 4 product before you know the subgroup into which the 5 patient will fall, and more of the patients that you 6 screen contribute data that informs clinically-7 important questions.

Agents should only be included in this 8 9 sort of approach when there is strong learn-phase 10 rationale, demonstrated penetration, PK, lack of 11 inactivation at the site -- that's the PTSD from the 12 lung example -- that makes it reasonable to test that 13 drug in that site against the likely spectrum of organisms. But with this approach, each enrolling 14 15 site -- meaning clinical study site, hospital, or 16 whatever -- can contribute a larger number of patients 17 per month because there are multiple infection types 18 and resistance patters included. And that decreases 19 the per-patient cost and helps you support your The efficiency of the platform is increased 20 network. 21 when there is more than one investigational agent 2.2 available at the same time for an indication because

of the shared control arm, but that is not necessary conducive to a platform strategy.

I have a couple of slides which I've 5 included in the site that have to do with labeling and 6 7 stewardship. The point simply stated is that when we 8 approve drugs that are most valuable from a public health point of view when their use is restricted to 9 10 highly-resistant organisms but we label them for broad 11 clinical indications, that that potentially undermines 12 stewardship efforts. I don't want to dwell on this, 13 because first of all, I'm not an expert in this area. But I do think that the statistical design drives the 14 15 data, the data drives the labeling, the labeling 16 drives the marketing, and we want to make sure that to 17 the extent possible that that supports stewardship 18 from a public health perspective.

19 So in conclusion, I think the most 20 common structure in statistical design of confirmatory 21 trials of antimicrobials risk failing to answer the 22 questions that are of most direct clinical urgency and

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1	impact. They fail to address the likely subsequent
2	use of approved products across multiple infection
3	sites, demonstrated only on the presence or risk of
4	highly-resistant pathogens and hopefully PK and
5	penetration data, but that a multi-infection site,
6	multi-drug platform trial addressing both non-
7	inferiority and superiority simultaneously could
8	address those challenges. Thank you very much.
9	DAN RUBIN: Thank you. Our next
10	speaker is Scott Evans. He is a Professor and
11	Founding Chair of the Department of Biostatistics and
12	Bioinformatics at George Washington University and the
13	Director of the George Washington Biostatistics
14	Center. He is also the Director of the Statistical
15	and Data Management Center for the Antibacterial
16	Resistance Leadership Group.
17	SCOTT EVANS: Thank you very much, Dan.
18	And good morning, everyone. Thank you for the
19	opportunity to talk with you today. I'm going to move
20	quickly.
21	A couple of years ago, I had a leaky
22	roof in my house. It created a water bubble in my

1	wall. It was a very strange-looking thing, but I had
2	a water bubble in my wall. And in addition to a new
3	roof, I had to repaper the wall. And my neighbor had
4	recently papered a similar-size room in his house.
5	And so I asked him how much paper did you buy. And he
6	replied six rolls.
7	Upon finishing papering the wall, I had
8	only used four rolls. And went to my neighbor and I
9	said, listen, I had two rolls left. What happened?
10	And he replied, oh, that happened to you, too?
11	Now, I tell you this story because I
12	asked the wrong question. And I find in clinical
13	trials we're often asking the wrong question as well.
14	And as a matter of fact, the two things I've learned
15	about antibiotic clinical trials since I've been
16	involved with them. First of all, they're rigorously
17	conducted by experts in the field, they closely adhere
18	to highest standards and fundamental principles of
19	randomized trials. And secondly, they're essentially
20	useless for helping clinicians make treatment
21	decisions. And so we've been working on ways to try
22	to figure out how to rectify this.

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1	This was said perhaps more eloquently
2	by the former FDA Commissioner Rob Califf. "Most
3	clinical trials fail to provide evidence needed to
4	inform medical decision-making. However, the
5	implications of this deficit are largely absent from
6	discourse."
7	So the example uses we have in
8	antibiotic trials is drugs are compared in susceptible
9	disease, but susceptibility is not known until we
10	actually start treating, after we've started treating
11	the patient. Patients are considered failure when
12	they change therapy, though they may not actually
13	fail. We lose interest in patients that change
14	therapy, despite therapeutic adjustments that could
15	effectively treat the patient. Populations studied
16	are not the same as the population applied. In non-
17	inferiority trials much of the time we exclude
18	patients with recent prior therapy. These drugs are
19	then used in these patients, possibly representing a
20	majority of these patients.
21	There are further issues. We often
າງ	define analyzig nervlations in trials . Efficiency

22 define analysis populations in trials. Efficacy

analysis we define an intention to treat population.
 Safety analysis we define as safety population. Those
 populations are not the same.

We then combine these two analyses into what we call a benefit-risk analysis. But to whom does this benefit-risk analysis apply? We're estimating a parameter from a population that doesn't exist. Nobody seems to mind.

Another question. We measure duration 9 10 of hospitalization, duration of the ICU stay in 11 clinical trials. Shorter duration is better. The 12 faster the patient dies, the shorter the duration. So 13 you give me a summary statistic of duration of hospitalization, I don't even know what it means. 14 15 Part of that's self-inflicted wounds by the way we design and analyze studies. Outcome interpretation 16 17 needs context of other outcomes for that same patient. 18 Once you tell me whether the patient lived or died, 19 the number makes sense.

20 Question three. Trials typically use 21 binary endpoints. For example, cure. The patient has 22 to survive, their symptoms resolve, they have

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1	microbiological eradication, no changes to therapy.
2	But consider the following. One patient fails because
3	they die. Another patient fails because they have a
4	lack of micro-eradication. It would seem reasonable
5	that now our primary analysis doesn't distinguish
6	these two patients. Shouldn't a primary analysis
7	recognize the difference of this? That would seem
8	important enough to recognize.
9	This would seem to be particularly

10 important recent FDA Advisory Committee for evaluating plazomicin in complicated UTI. The composite cure 11 12 rate, 81 percent for plazomicin, 70 percent for 13 meropenem. But if you just look at the clinical cure rate, very close. Right around 89, 90 percent for 14 both. It turns out that the advantage for plazomicin 15 16 is in the micro-eradication. That would be 17 particularly important to know particularly when you 18 start to look at the safety data that suggests, well, 19 there's more safety concerns with plazomicin than 20 perhaps with meropenem.

The last lesson I'd like to sort of
motivate this is one plus two times three is not nine.

1	And grade school children know this, but the clinical
2	trial community have missed this course. Let me
3	explain what I mean by this. So here's a question for
4	you. Supposed a loved one is diagnosed with a
5	terrible infectious disease. You get to elect
6	treatment. We have three treatment options; A, B, and
7	C. All right?
8	Now, let's suppose there's two outcomes
9	for simplicity. Let's assume they're equally
10	important. You have treatment efficacy, treatment
11	success outcome. Yes or no. The patient gets it or
12	they don't. They also have a safety event. Patient
13	experiences it, yes or no. Now, luckily enough, we
14	had a randomized trial that compares A, B, and C that
15	help guide our decision about which one we should
16	choose. Had a hundred patients in each arm, A, B, and
17	C. The treatment success rate is 50 percent in A, 50
18	percent in B, and 50 percent in C. Now, the safety
19	event rate is 30 percent in A, 50 percent in B and C.
20	Which treatment do you want?
21	Well, they all have the same success
22	rate. A has got the lowest safety event rate. B and

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C are indistinguishable. Can't tell the difference
 between them. Clearly, we choose A. We're all
 reasonable people.

4 But instead, what we've done here if 5 you evaluate what we've done here, we've taken the patients in the trial and analyzed the outcomes, the 6 two outcomes. What I'd like to do is flip that upside 7 8 down. Take the outcomes in the trial and tell me what happened to the patient; that the purpose of measuring 9 10 outcomes in clinical trials is to tell you how the 11 patients are doing. We seem to have gotten it 12 backwards.

So if I take the outcomes and analyze what happens to the patients, there are four possible outcomes for what happens to patients. They get the treatment efficacy, yes or no, and they get the safety problem, yes or no, in combination. So let's look at the combination.

Well, it turns out in treatment A, the efficacy and the safety were uncorrelated. So there were 35 patients that experienced the treatment success and avoided the safety problem. But in

1 treatment B, they were positively correlated. So 2 there's zero patients that had treatment efficacy without the safety problem. In treatment C, they're 3 4 negatively correlated. So I've got 50 patients that 5 experienced the efficacy without the safety outcome. 6 Now, one slide ago I couldn't tell you 7 the difference between B and C. Now, we're supposed 8 to be in an area of personalized medicine where we're doing personalized -- and I can't tell the difference 9

10 between B and C. So our culture has been to use the 11 patients to analyze the outcomes. Shouldn't we use 12 the outcomes to analyze the patients? That's the 13 purpose.

So as my father told me many years ago, the order of operations is important, and we haven't got the order right. And we may be missing things without realizing it.

So William Osler, the well-known
clinician, said years ago, "The good physician treats
the disease. The great physician treats the patient."
And maybe we should be analyzing things that way.
So Dean Follmann at NIAID and I wrote a

1	paper describing some of these issues and how we might
2	attack things from a different angle in doing so. A
3	couple of years ago we wrote this paper, DOOR,
4	Desirability of Outcome Ranking. Think about what
5	those words mean; desirability of the outcome. And
б	it's going to be a ranking. And what we do is we end
7	up computing what we call the DOOR probability, which
8	is a probability of a more-desirable outcome when
9	assigned to one therapy relative to another therapy.
10	Now, if you're a clinician treating a
11	patient or you're a patient and asking I've got to
12	treatment options, you can ask for differences in
13	means and difference in proportions and hazard ratios
14	and relative risk all you want. But what is a more
15	natural question to ask when I'm trying to figure out
16	whether I should take one therapy over another? Well,
17	how about the global probability that one therapy is
18	better than another? Wouldn't this be intuitive? So
19	although it may be foreign at the moment, that perhaps
20	this may be intuitive in the long run.
21	So here's an example of an application
22	of this. Should we use ceftazidime, avibactam, or

1	colistin for the initial treatment of infections due
2	to CRE? And this was a paper published by David van
3	Duin using some observational data from an ARLG study.
4	Now, the DOOR that was set up, a
5	desirability of outcome ranking, an ordinal outcome.
6	You see most desirable at the top. The patient lives,
7	they're discharged home, everything's back to normal,
8	they avoided major adverse events. Least desirable is
9	at the bottom; the patient dies. But there's layers
10	in between where some things go right, but not
11	everything goes right. And the idea is if colistin,
12	whatever colistin produces, patients fall into these
13	categories wherever they fall. Can I get patients to
14	migrate northward using CAZ-AVI to more desirable
15	places and evaluate that. And that's exactly what
16	happened in this particular evaluation. The DOOR
17	probability is 64 percent that you're going to have a
18	better outcome using CAZ-AVI than you are in colistin.
19	Now, how do you summarize this sort of
20	patient journey, which is the next question if you're
21	going to use DOOR? Now, before analyzing several
22	hundred patients, you have to figure out how you're

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1	going to analyze one. Now, clinicians are supposed to
2	be doing this when they're treating patients, but
3	we've got to get into the head a little bit more.
4	So there was an example strategy about
5	how we might create this ordinal outcome. This was
6	led by Sarah Doernberg at UCSF through another ARLG
7	study. And we call this BAC DOOR, because this was
8	for bacteremia. And what we did was we were
9	envisioning we're going to do a staph aureus
10	bacteremia trial. And we did a pretrial sub-study to
11	try to figure out how we might create a DOOR in this
12	particular area.
13	So what we did is we took 20
14	representative patient profiles from a prior study
15	that had benefits and harms and quality of life, and
16	in a single paragraph we wrote down what happened to
17	those 20 patients.
18	We then sent those profiles to 43
19	expert clinicians that treat this disease and said
20	rank them in terms of the desirability of their
21	outcome. We didn't tell them how to rank them; the
22	idea was to figure out what are they valuing in terms

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1	of how they rank patients.
2	And then we examined the components
3	that drive the clinician rankings and used that to try
4	to create a DOOR outcome. And we came up in this
5	particular case with six different levels.
6	Now, one thing we learned from this
7	particular exercise. First of all, you for the first
8	time begin to evaluate the cumulative nature of things
9	that happen on patients. Now, fi you examine one
10	outcome at a time, you never see anything cumulative.
11	We never look at it. But if a patient is having
12	multiple bad things happening to them, then that
13	should be recognized. That's how patients are
14	experiencing these outcomes, and that we should be
15	thinking about how to do that.
16	So some natural questions arise when
17	you evaluate things like this. There are potentially
18	unequal steps between these categories. We like to be
19	able to recognize that. Perhaps the step to the
20	bottom category, to mortality, is bigger than the
21	steps above it, is larger than the steps above it. We
22	can recognize that. There's also varying perspectives

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1	among patients and clinicians regarding the
2	desirability of these different levels, and could we
3	recognize that. And we have done this sort of
4	analysis. We proposed what's called a partial credit
5	analysis, which if you've got these ordinal outcomes -
6	- these are the four ordinal outcomes from the CAZ-AVI
7	colistin example what you do is you say, well, if
8	you have the most desirable outcome, you get a perfect
9	score; you get a hundred. If you die, you get a zero.
10	If you get in the middle, then you get partial credit.
11	And there are some very natural ideas about how we
12	next natural question is what are you going to give me
13	for partial credit. And we've been working on that
14	particular problem and we've got some ideas for that.
15	The other thing that can happen with
16	this is there's been this debate in infectious
17	disease, this area of infectious disease for quite
18	some time that says, well, who do I want to enroll in
19	trials. And one theory says, well, don't enroll the
20	very sick patients, because they're going to die
21	anyway and you're not going to have any sensitivity to
22	detect any sort of effects.

1	On the other hand, people say, well,
2	don't enroll the very more healthy patients, because
3	they're going to recover anyway, and you're not going
4	to detect any effects, there's not going to be any
5	sensitivity. Well, in this particular case if you
6	look at things using a DOOR type outcome, you can
7	evaluate which patients are actually benefiting from
8	say CAZ-AVI over colistin. And in this particular
9	case, the most severe patients were the ones who were
10	benefiting over CAZ-AVI CAZ-AVI over colistin.
11	Let me show you this idea in a little
12	bit in another example. This is the PROVIDE study,
13	another ARLG study, which was a prospective, multi-
14	center, observational study evaluating among
15	hospitalized patients with MRSA bloodstream
16	infections.
17	The research question; what's the
18	vancomycin PK exposure target that's associated with
19	an optimal treatment outcome. All right? 265
20	patients in this study.
21	Now, what we did is we set up a DOOR
22	outcome. Most desirable at the top says treatment

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1	success, and you avoid major toxicity, acute kidney
2	injury. Least desirable at the bottom is the patient
3	dies. But there are gradations of patient response
4	along the way.
5	So what we did is looked at DOOR
6	outcomes by dosing quintiles. So the top dose there
7	is the highest bar. And what you see here is a
8	distribution of the DOOR outcomes by dosing quintiles.
9	Highest dose at the top, lowest dose on the bottom.
10	All right?
11	So the blue on the left-hand side is
12	that most-desirable category, treatment success
13	without acute kidney injury. The purple on the far
14	right is mortality. So what are you getting as you
15	increase dosing? What you get is toxicity. You're
16	not getting more efficacy; you're getting more
17	toxicity. So higher doses bring higher toxicity, but
18	not greater treatment success. All right. So
19	actually this particular paper came out today in the
20	Annals of Internal Medicine using ideas like this to
21	actually monitor patients during the course of say DMC
22	monitoring.

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1	So in ARLG 2.0 one thing we're hoping
2	to do is to develop standardized, syndrome-specific
3	DOORs for the major infections in this particular
4	area.
5	My last point I would like to make.
6	There's been some talk today about platform trials and
7	so forth. And in the ARLG we actually evaluated a
8	platform to try to get at more pragmatic answers to
9	the questions that are most important for us. We
10	called this SMART-COMPASS. And the idea is so
11	SMART-COMPASS stands for Sequential, Multiple-
12	Assignment, Randomized Trials for Comparing
13	Personalized Antibiotic Strategies. So this is
14	consistent with the strategy theme that Vance and
15	others had talked about earlier. But it's actually
16	quite flexible and very consistent with the way in
17	which patients are treated. It addresses several
18	types of research questions, including identifying
19	optimal treatment strategies, it can evaluate empiric
20	therapies, and it can evaluate definitive therapies,
21	those that are likely to be licensure-type questions,
22	and provide efficiencies compared to traditional

1 multi-arm trials. And it's very pragmatic in the sense that it really mirrors clinical decision-making. 2 You can think of it as personalized medicine. 3 4 So I'm going to end with a quote from 5 NBA coach Frank Layden, who had a player that was not producing. And Layden asked the player, so what is it 6 7 with you, son? Is it ignorance or is it apathy? And 8 the player looked at Layden and said, Coach, I don't know and I don't care. 9 10 So I say to you today that if people

don't know, let's educate them. And if they don't care, then let's motivate them. And I want to thank my collaborators, Dean Follmann at NIAID, Dan, and Chip and Vance from the ARLG, and all the ARLG team. I have no doubt that you will enthusiastically applaud now because you're so relieved that it's all over. Thank you.

JANE KNISLEY: Okay. Our next speaker is Aaron Dane. He is a consultant statistician who has been investigating how to make the design and interpretation of antibiotic trials more feasible. He has been a consultant since April 2016 and has

1	continued to make such development more feasible.
2	AARON DANE: Thank you. So as has just
3	been said, all I've been looking at is for a while now
4	is seeing more and more that we talk about doing rare
5	pathogen trials. So this is a bit different from some
6	of the other presenters in that this is in that area
7	where it's very hard to get very much information.
8	But we seem to be stuck in that if we can get 50 to
9	100 patients with a resistant pathogen in a pretty
10	long timeframe, we've got nothing we can do with the
11	data. Because we can't get to traditional statistical
12	criteria so we run the trial and then we can't do
13	anything with it.
14	So as a result of that, I've been doing
15	some work with Professor Nigel Stallard at Warwick
16	University where we've been looking at and he's
17	taking ideas from the orphan drug area where how can
18	we use that information a bit more readily and get
19	something from the data, particularly in areas where
20	very specifically we can't generate much more
21	information.
22	So we have a technical problem.

Page 108 Well, maybe while they're sorting it 1 out what I'll do is just -- so what I'm doing here --2 the idea is can we get more from a small study when we 3 4 haven't got the option of a bigger study? And what I'm looking at here is very specifically we're still 5 talking about doing a randomized trial. So this is 6 7 still randomized, it's still controlled. This isn't 8 about external controls, but it's having done that because we still feel that's the best way to get some 9 10 useful information; what can we do with the 11 information we have got? Because as we talked about 12 yesterday, at the moment we do the study and then no 13 one can use the information. So it's just a way of doing something a bit differently. 14 15 Okay. Is it working? All right, okay. 16 Try number two. 17 So these are my disclosures. And so 18 this was a collaboration with Nigel Stallard in 19 Warwick. He has done some work in the orphan drug So basically that was the idea here, was to try 20 area. 21 and take some of that thinking when you've got that 2.2 limited population and see if we could apply some of

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1	that. And Paul Newell and John Rex have also helped
2	as we've worked that through to put some clinical
3	perspective on that as well.
4	So I'll skip through this quickly. So
5	in terms of the superiority studies, this has been
6	covered a number of times. What I would say is the
7	idea is if we could do superiority studies, we would.
8	It makes things much easier and much clearer. So it's
9	not that we'd rather do non-inferiority studies, it's
10	just that they often not really stick. For the
11	reasons that have been spoken about, we can't study
12	ineffective comparators. The numbers are small just
13	by definition, which makes it more challenging. And
14	there's often at least one therapy with some degree of
15	efficacy. It may be toxic, there may be various other
16	reasons you don't want to use it. But it's not that
17	we're often dealing with something that's completely
18	ineffective.
19	And also from the societal point of
20	view, we don't want to just be developing new
21	antibiotics when we can show superiority; we want to
22	do it before that time.

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1	So this was really just to illustrate
2	how quickly the numbers fall away when we're talking
3	about resistant pathogen studies. This is just an
4	example, it's made up. But it's this illustration
5	that from 300 patients if you're trying to then focus
6	on a very specific pathogen, your numbers fall away
7	very quickly. And that's before you even talk about
8	that pathogen being resistant to all other therapies
9	and all the other comorbidities that may cause
10	confusion. So this is why we still need to find ways
11	of developing trials for non-inferiority setting as
12	well as superiority, even when we're talking about
13	resistant pathogens.
14	And I think the key thing is that a lot
15	of the time we're developing this for tomorrow's
16	patients. So it's for a future unmet need as much as a
17	current unmet need. And that's why we have to do it
18	this way.
19	And I think then what we need is find a
20	way of saying, well, how can we develop something when
21	we can only get a hundred patients? So what do we do
22	with that information rather than saying, well, that

doesn't meet traditional statistical criteria, so we 1 2 just can't do anything. So it's finding a way of getting some information -- the conversation yesterday 3 4 was, well, these drugs are still going to be used, so it must be better that we provide some framework for 5 whether that drug's effective even if it's not the way 6 7 we would do that traditionally. Something I'm not going to mention in 8 this talk is the idea of the safety database. 9 It's 10 obviously an important element as well. And that 11 would have to be considered. But I'm focusing on the 12 efficacy aspect for this talk, but that would clearly 13 be an important aspect as well. So a possible approach to design of 14 15 rare pathogens. So some of this is standard for any trial; what are we interested in when we run a trial? 16 17 So we want to be confident that we can show an 18 effective treatment works, and we want to be confident 19 that we won't approve ineffective treatments. That's the premise of any trials, and that's a Type I error 20 21 in power. 2.2 But the question is can we look at that

differently for rare pathogens and can we do that for example using some of the questions in the orphan drug area about understanding what extra information and value we gain from making the study bigger.

I think one of the key factors for me 5 here is -- because often this discussion is saying, 6 7 okay, it would be acceptable to do a small study, and that's all right. But actually we still need a 8 framework for the decision-making. So as a sponsor, 9 10 you need to understand what you need to show so you 11 can understand the risks. These studies may be small, 12 but they take a long time and they're very expensive. 13 So we still need to understand how likely we are to be successful. And also just having clarity at the 14 15 outset on what the decision-making is going to be means that it's much clearer to everybody what's 16 17 needed.

So the aim here was to provide a framework in that setting where you've got other therapies that may be suboptimal, but they still have some efficacy. And what I'm going to go through, I'm going to step through some examples. And this is not

1	about performing an interim analysis or not about
2	continuing on after the trial in a single-arm setting.
3	What this is about is saying if you've got a limited
4	population, you run a trial in part of that
5	population, and then everybody else goes on to one of
6	the two treatments. Either they go on to the test if
7	it had a successful outcome, or they continue on the
8	standard of care if your new treatment failed in the
9	study.
10	The other thing, I've stuck to
11	frequentist statistics here purely from the point of
12	view that I'm talking about a different idea anyway,
13	so I didn't want to get into the Bayesian side as
14	well. But you could equally apply this with Bayesian
15	methods, some of the things I'm looking at.
16	So the first question is large versus
17	small trials with rare pathogens. So larger trials
18	lead to higher power, so that's great. Right? You'd
19	always want that. But what we have here is that if
20	the trial is too large or it takes too long, what it
21	deprives patients of a more effective therapy, and it
22	also means it might not be feasible to develop the

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1	drug in the first place.
2	But on the flip side, if the trial was
3	too small, it may be more feasible, but then you end
4	up with much higher chances that you're going to make
5	the wrong decision, which clearly we don't want to do,
6	either.
7	But the theme for all of this is how do
8	we work with a small data set? Because that's all
9	that's going to be possible. We're not going to be
10	able to generate any more. And that part isn't really
11	much of a choice. So can we show there's a sweet spot
12	for the sample size where we get sufficient
13	information and we're not going to gain a lot more if
14	we keep going on with the trial? And really that's
15	the idea of the diminishing returns outside of that
16	sweet spot.
17	So what we're aiming for here, so if
18	test is worse than control, then in that situation,
19	every patient randomized to test in a trial risks a
20	worst outcome. So if test is approved, we perpetuate
21	that problem, and there's even more patients that are
22	having a worse outcome than they would otherwise. So

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1	the mitigation here is that we have a small trial, but
2	we avoid or reduce that chance of incorrect approval,
3	which is what we normally talk about as Type I error.
4	Equally, if a test is better than
5	control, then every patient randomized to control
6	risks a worst outcome. So in this case if test is not
7	approved, then that's perpetuated. So you're still
8	risking everybody having a worse outcome. So here
9	within that small trial we want to keep the power
10	high.
11	And then the third scenario I'll
12	present is when test is similar to control. So in
13	this case, we still want to make additional therapies
14	available. So again, in this case within that small
15	trial, we'd want to keep the power high.
16	The important part with all of this
17	though, this all sounds fine and easy. But the fact
18	is we don't know which of these three scenarios is
19	true when we're designing a trial and when we're
20	interpreting a trial. So this is where we need to
21	understand the risks we have of all these things
22	happening. So what are the risks that we incorrectly

approve a new drug, or what are the risks we miss a
 good drug by the criteria we apply?

3 So what this is presenting -- so I'm 4 going to present three different scenarios, one for 5 test being better, one for it being worse, and another 6 one where it's the same to understand some of those 7 risks.

8 So in this situation imagine that we've got a non-inferiority trial which is looking at test 9 10 against control. The test response rate is 60 percent 11 and the control is 40 percent. So what we've chosen 12 is a non-inferiority margin of 20 percent here with a 13 95 percent confidence interval. So we've already started with a wide margin because of the rare 14 15 pathogen and the area of unmet need. So the correct 16 outcome and what we want to say in this case is that 17 we conclude non-inferiority.

So we run the trial. And the plot on the right shows the chances of concluding noninferiority. And what it's showing is that from about 40 or 50 patients per group, you've got a reasonable chance that you're going to show non-inferiority. So

1 80, 90 percent power. So you might need more for the 2 safety database, but in this scenario what that's 3 showing you is that you'd have a decent chance of 4 success given that situation.

And then on the flip side, if the test 5 was 20 percent worse than control, so if the response 6 7 rates here were 20 percent and 40 percent, what would 8 happen then? So we're still using the same NI margin and confidence interval, but in this case you don't 9 10 want to conclude non-inferiority. And what that's 11 showing is that the chance of concluding non-12 inferiority on the plot now is very low. So in this 13 case, again, 40 or 50 per group would be okay for this situation. Because what it's showing is you rarely 14 15 make the error of concluding non-inferiority.

And finally, when the test and control are the same, we have all the same setup. So this time they've both got a 40 percent response rate. Now, as you would expect, as the sample size gets bigger, the chances of concluding non-inferiority goes up. But what this is showing is that it would take a bigger study in this case to do that. So at 50 per

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1	group, you'd only have a 50/50 chance of success even
2	when you're actually truly the same as the comparator.
3	So around 80 per group may give you
4	what you need. And it takes a long time to get beyond
5	that. But I think the key question here is you'd need
6	a substantial bigger study, and otherwise, you'd end
7	up with a 50/50 chance for success.
8	And the reason I raise this is in some
9	situations, that might be fine. It might be quite
10	easy to get to 200 patients in one of these studies.
11	But if it's not and you could only realistically get
12	to about a hundred, a sponsor isn't going to put the
13	time and money into a study if they've only got a
14	50/50 chance of it succeeding.
15	So what I've gone through so far is
16	sort of the more standard aspects of Type I error in
17	power. But the other part that's been pulled through
18	from the orphan drug area is what does this mean for
19	patients after the trial. So in this case, suppose we
20	had an overall population of patients with a rare
21	pathogen. There's a thousand of them in total. And
22	then if we include a hundred of patients in the

1	clinical trial, so 50 per arm, that means we treat the
2	remaining 900 patients with whatever drug comes out of
3	that study. So that means we have 50 patients on
4	test, 50 on control, and then another 900 on one or
5	other of the drugs depending on whether the trial
6	concluded that the test was non-inferior or not.
7	And the purpose of showing this is that

/ 8 a bigger trial isn't always better with rare diseases when you're doing this. So if we assume we had a 9 10 thousand patients with the rare pathogen and we had a 11 hundred patients in the randomized trial -- and from 12 what we looked at before, if we assume that we have a 13 60 percent response rate for the test and 40 on control, as I showed before, what that means is that 14 15 you would expect 30 of your 50 responses in the trial 16 on test, and 20 out of 50 on control to respond. So 17 that's the 40 and the 60 percent response rates.

And then dependent on whether that trial concluded non-inferiority or not -- so if the test was successful and got carried through, from that point, those remaining 900 patients you would expect to have 60 percent of them to respond. Okay? Whereas

1 if test failed, so you didn't conclude non-2 inferiority, that would mean that the standard of care would continue to be the treatment used. And so only 3 4 40 percent of those 900 patients would now have a 5 response. And I won't get into all the details of 6 7 the power and the probabilities of those two. But 8 what you can then work out is the expected number that you expect to respond, which in this case would be 587 9 10 of the thousand. So that's a 58.7 percent response. 11 So that's fine. 12 But now then you'd say, okay, why don't 13 we just do a bigger study? Because then we'd get more certainty that we have the right decision. But for 14 15 this situation and this scenario, if you did the same thing with a trial of 400 patients, what happens now 16 17 is you've still got the same 60 percent response on 18 test and 40 on control, only with 200 patients per arm 19 in the study now. And what that means is overall there's only 560 patients, or 56 percent of all the 20 21 patients with this pathogen respond. And the reason 2.2 for that is because you're waiting longer and you're

giving more patients the ineffective therapy before you switch to one therapy or another. So this is where the consequences can be worse for patients afterwards if this is the case.

5 So in terms of finding the sweet spot that I mentioned before, what we need to do is find a 6 7 sample size we have a good chance of success when we're effective, low chance of approval when we're 8 ineffective, and a reasonable chance of success when 9 10 the two treatments are similar. But also consider 11 this expected number of patients who benefit so that 12 we maximize that. So the following plot summarizes 13 that information.

So this is plotting out that expected 14 15 chance of success. So here the top half is what I presented earlier, which is to say how often do we 16 17 select the test agent. So when test is 20 percent 18 better, we pick that pretty well from 40 per arm. And 19 what the bottom plot shows is that a trial with 40 patients, you ought to optimize that number of 20 21 successes when you get to about 40 or 50 patients. 2.2 And after that, it's diminishing returns because

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you're keeping more patients in the study even when 1 one of the treatments is less effective. So this is 2 3 where you can then use this to look at that and say, 4 okay, well, maybe 40 or 50 patients is reasonable. 5 Similarly, when test is 20 percent 6 worse, what happens here is -- we already mentioned 7 that the test is not selected very often. And again, 8 the expected number of patients who respond drops right from the start in this case. And that's because 9 10 in this case you run the study longer, and that means 11 you're continuing to treat more patients on the test 12 agent even when it's not effective. 13 So finally, the largest sample size for test is similar, as I mentioned before. 14 So that goes 15 up slowly. And you'd need to get to about 80 or a hundred patients pre arm before you'd get to a 16 17 reasonable chance of success. In this case, the 18 number you'd expect to respond stays the same. And 19 that's because both treatments have got the same level of efficacy. So whichever one you choose, your 20 21 expected response would still be 40 percent in this 2.2 case.

1	So the reason I've gone through all of
2	that is because what this shows is that 40 to 80
3	patients per arm has reasonable power in these
4	settings that I've shown. And that's assuming that
5	the control has 40 percent efficacy and then the test
6	has either 20 percent better or worse response.
7	But the fact is if the product has
8	similar efficacy to the control at the moment, that
9	needs more patients, a bigger study, which in some
10	settings might be possible. But from what we've been
11	talking about, that's not always going to be possible.
12	So getting to 200 patients with rare pathogens is not
13	always the case.
14	So if that's the case, how could we
15	provide criteria when it's only feasible to recruit 50
16	patients per arm, for example? And that could well be
17	taking five years plus to get to that.
18	So one approach then would be to look
19	at the plots that I've just mentioned, but use
20	something like 80 percent confidence intervals rather
21	than the traditional 95 percent confidence intervals.
22	Now, what I would specify here is that

1 this is in areas of large unmet need, and it would be 2 very specific. So this wouldn't be a blanket approach 3 to this. They would have to be specifically agreed 4 that this is the idea that we understand the risk, but 5 we do it in situations where the alternative is we 6 have no data.

7 So I won't go through these plots in I'll save you all that pain. But what 8 detail again. I'll summarize is that what shifts here is the 9 10 observations that in this case is that when test is 20 11 percent worse, the risk of approval goes up, the 12 incorrect approval. So now it's about a ten percent 13 chance that you'd conclude non-inferiority for a lesseffective treatment. But when the test is similar or 14 15 better, using the 80 percent confidence interval gives 16 you a higher chance of success than using the 95 percent confidence interval. And the pattern for the 17 18 expected number of responses is similar to before. 19 So the important thing here is that this is example framework. So I'll give an example of 20 21 using test and control being 20 percent better, 20 2.2 percent worse. And I've given examples of 80 percent

confidence intervals. But the idea is this is a 1 2 framework to say at the moment what we do is we say, well, we run a resistant pathogen study, and if it 3 4 doesn't meet standard criteria, then we can't use it, and we know we can't get to those criteria. So this 5 is a way of discussing what the unmet need is, what 6 7 the feasibility is, and trying to get some idea of where we're going to be with some of these aspects and 8 change some of the success criteria that we might have 9 10 for these very specific situations.

11 So in summary, this is a framework to 12 display the tradeoffs. So this is a situation when 13 only a small trial is possible. So we need to understand the false positive and negative error 14 15 rates, but we should think about whether we can change 16 what those error rates are compared to a traditional 17 area when we can study hundreds of patients. And the 18 data on 100 to 200 patients can be very informative. 19 And it feels like it's still better to generate that 20 data and understand what it's telling us even if we 21 have to acknowledge that the levels of risk and the 2.2 level of information is not the same as we are

1	traditionally used to.
2	So in conclusion, I would say that the
3	use of the power Type I error and the overall number
4	of patients benefitting could be used to agree the
5	sorts of criteria for these treatments of rare
6	pathogens so that we get some information where
7	everybody understands what we're going to see as a
8	successful trial at the outset rather than us having
9	to do the trials and then not really knowing what
10	success is going to be until we get to the end of the
11	study. Okay, thank you.
12	DAN RUBIN: Thanks, Aaron. We now have
13	scheduled 15 minutes of Q&A and we'll go until 11:20,
14	and I'd like to open up the Q&A for speakers for both
15	sessions from this morning. If you'd like to make a
16	comment, please just turn your card up, as was done
17	yesterday. Try to get my attention or Jan's attention
18	and we'll try to keep track of the order. Please, go
19	ahead.
20	MANOS PERROS: Thank you. Just a
21	comment and then a question. I find it shocking but
22	not surprising that I believe, Vance, 87 percent of

the prescriptions in your institution for those new drugs were off-label. It's shocking. And I think we all need to decide whether we can settle for something like that or whether we want the labels to be more reflective of how the drugs are actually used. I would be in the latter camp but that's a discussion to have.

On the more specifics, I like the idea 8 of registration trials, strategic trials. 9 Let's 10 remember that we also have other groups than 11 prescribers and regulators that need to be satisfied 12 in order for us to be in this business, and that 13 includes pharmacists, P&T committees, payers, and who is going to -- what kind of data and what kind of 14 15 information is (indiscernible) those groups.

16 And the bottom line of that is more 17 trials take time and money that we don't have. Maybe 18 Merck could do that. I don't know why they would but 19 they could run. But we only launch a drug once. The price is set at launch. The trajectory in the first 20 21 quarters is important. And while I love the idea, I 2.2 think we need to think how we can do those kind of

1	trials before the drugs I launched.
2	VANCE FOWLER: Well, yeah, we can think
3	about that all the you know, all the livelong day.
4	I think that I mean, I think you You know,
5	what's that saying about don't let the perfect be the
б	enemy of the good. I think we simply what I'm
7	thinking about from a clinical standpoint is we need
8	these drugs. You know, we need them today. And that
9	87 percent, that's that is essentially if
10	anything, it's maybe a little low. I mean, that's
11	been Mike Rybak published you know, you talk
12	about ceftaroline a little bit more, for an example.
13	So, we've been you know, that
14	compound, incredible compound in my opinion, you know,
15	it's been handed off it's been owned by no less
16	than five different entities so far. So, Cerexa, then
17	it was Faris, then it was something that started with
18	an A, then it was Allergan. And now Allergan's going
19	away. I mean, so, you know and the likelihood of
20	that of meaningful trials taking place with each of
21	those handoffs plummets.
22	You know, Mike Rybak published a thing

1	from 2014, that's the best data we have to deal with
2	it. It was a retrospective series of, I don't know,
3	about 400 patients. Over 80 percent of those patients
4	he described, and that was in 2015, were off-label.
5	So, that's the status I mean, that's
6	the state of affairs in clinical care, treating drug-
7	resistant pathogens in the United States. That's just
8	the way it is. And so I don't see that, you know
9	And if we can go in and do trials beforehand, totally
10	agree. I don't care when the trials get done,
11	honestly, from a clinical standpoint, as long as the
12	clinicians have the you know, can confuse the issue
13	with facts about what we're actually doing, it's a win
14	for the patients, right?
15	So, I'm just you know, what I'm
16	trying to really propose is I guess maybe I'm saying I
17	don't think it is going to happen, you know? Maybe
18	I'm a pessimist. I usually am. I think that what can
19	happen is we can get, you know that companies can
20	get compounds through trials and complicated UTI and
21	ABSSI or what have you, because it minimizes risk.
22	And then what I would like to see

1	happen in the next ten years, I do think clinical
2	trial networks are the way forward. I feel what I
3	would like to see is a means by which there that
4	funding can be provided not just even from a single
5	country but on a multinational perspective whereby
6	everyone puts in a little bit, there's a means by
7	which decisions responsible implementation of that
8	precious resource can be applied to do the trials that
9	have to be done.
10	So, yeah, maybe that's a little you
11	know, big but, you know, someone's got to think big,
12	you know, or we're just going to stay right where we
13	are. It's the clinical it's the clinical community
14	that's going to have to take on these tough-to-treat
15	trials, you know, tough-to-complete trials because it
16	just doesn't You know, you guys have already got
17	so much stuff just trying to stay alive that adding a
18	putting your entire future on a single trial that
19	is too that's never been done before, to me, is a
20	bridge too far.
21	MANOS PERROS: Yeah. And, Vance, just
22	to be clear, my point is not that I don't disagree

1 with anything you said. I'm not arguing that we
2 should change the way those drugs are used. This is
3 the right way to use them. But the point I'm trying
4 to make is getting the drug approved is no longer
5 enough for us to survive as an industry.

So, we have a question 6 DAN RUBIN: 7 online from John Tamico -- from John Tamico to Roger We'll do that and then I've got Dr. Rex. 8 Lewis. So, the question, Roger, is "CIAI is partly a surgical 9 10 disease, even resistant organisms can be managed with 11 source control to the point host defense can clear 12 infection confounding antibiotic effect. How is CIAI 13 reliably informative across body site analysis?"

Okay, well, as a non-14 ROGER LEWIS: 15 expert in this area, I think I'm going to take this as 16 the general question of if there is reason to believe that a treatment is likely to be less effective in one 17 18 setting than another, then that violates an assumption 19 of exchangeability of the body site. So, one of the assumptions of a hierarchical model is that a priori, 20 21 you're not sure which site the treatment effect is 2.2 likely to be largest if you couldn't order them a

1	priori.
2	So, if you have reason to believe that
3	an antibiotic is likely to be relatively less
4	effective because of other things that affect outcome,
5	then you have to adjust for that in some way. And
6	perhaps pooling it doesn't make sense.
7	That said, what I would want to do is
8	look at trials of treatments of antibiotics for
9	complex intra-abdominal infection in which the
10	isolates were sensitive and look at the magnitudes of
11	the treatment effects based on the endpoints that were
12	used, which presumably were compromised by the same
13	effects, and see if it's really true that the
14	treatment effect that we have observed form previously
15	demonstrated efficacious antibiotics was really
16	smaller to make that assessment.
17	DAN RUBIN: Thank you. Dr. Rex?
18	JOHN REX: Those were great talks. And
19	back to you, Roger, this Stein-Lewis concept is almost
20	causing my head to explode. And if I say it back to
21	you it's that if I have a point estimate of some
22	measurement, and then I have three point estimates of

three measurements, I'm actually -- the most valuable measure of my first point estimate isn't the actual number; it's actually a combination of the three point estimates. That seems to be the concept here that I can -- and the error in one direction might compensate for error in another direction, if I'm catching the drift of this.

8 And it seems to me that has some very interesting translations for us when we use -- you 9 10 know, we focus on the one-off data -- that's my point 11 estimate in CUTI. That's my... And it almost seems 12 to suggest we're better off doing studies where we 13 actually deliberately grab several different body sites as part of it and get -- and combine them. Or 14 15 am I misunderstanding this? Talk -- talk a little bit more about James, Stein and their very 16

17 counterintuitive concept.

18 ROGER LEWIS: So, the first has to do 19 with understanding the difference between bias and 20 estimation in treatment effect. So, the James-Stein 21 theorem says that if what you're interested in is the 22 error in estimating the treatment effect -- and I mean

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1	error like mean squared error you get a lower mean
2	squared error if you use information from all of the
3	related things.
4	The it is not an unbiased estimate,
5	okay, but the bias what you lose in the
6	unbiasedness, in my view, in terms of the clinical
7	utility estimate, is far outweighed by the benefit you
8	get in the reduced noise that Dan mentioned.
9	So, the and you can think about it
10	in the following way: Let's suppose you do a clinical
11	trial and for a treatment in osteomyelitis, and
12	there is a forest plot presented which shows the
13	estimated treatment effect across all the different
14	bones. You can't get to 206 but some large number of
15	bones.
16	There is going to be some bone where
17	the estimated treatment effect is surprisingly high or
18	surprisingly low compared to all the other bones. And
19	you as a clinician will look at that and know that
20	that bone might be a little harder or easier to treat
21	but it's probably not as much harder or easier to
22	treat as the data suggests because there is the the

1	estimates that are particularly high and low are
2	likely to be a combination of the true treatment
3	effect being high or low and random fluctuation.
4	And what the James-Stein estimate or a
5	shrinkage estimator does is it accounts for the
6	expected variability in multiple measurements to give
7	you an estimate that appropriately balances the
8	evidence that one bone or subgroup is different than
9	the others against the additional likely variability.
10	That's what the shrinkage estimate does. When it does
11	that, the distance between the estimated treatment
12	effect and the truth, on average, is smaller.
13	So, what this means and I know
14	there's a lot of people who do editorial work here
15	is that if you look at the forest plot or the table of
16	subgroup effects in a clinical trial, that is
17	fundamentally the wrong estimate if your goal is to
18	give you the most accurate treatment effect estimate
19	in each subgroup. It's fundamentally wrong.
20	LINDSEY BADEN: Yeah, but that's how we
21	do it. I mean that's so interesting. Wow.
22	ROGER LEWIS: I wasn't debating what's

Page 136 1 in print. 2 MAN 1: But that has its own 3 assumptions. 4 LINDSEY BADEN: But people spin around 5 So, the interesting thing 6 ROGER LEWIS: 7 about it is that the assumptions are extremely loose. 8 And so the reason that this is not adopted as a standard approach I think has to do with a limitation 9 10 in humans to understand situations where statistics 11 work in a way that's counterintuitive. Just as if 12 most people's experience doesn't include the fact that 13 time slows down when things go really fast, okay? Ιt just turns out to be true. 14 15 I would just add on the DAN RUBIN: James-Stein theory -- that the theory is that your 16 17 mean squared error must improve if there's three or 18 more subgroups. If you're talking about your average 19 error or the sum of the errors across the groups, if there's any, you know, one outlying group, there's no 20 21 guarantee that that, you know, within subgroup 2.2 estimation must improve by bringing in other data --

Page 137 although, you're right, that on average that shrinkage 1 2 is going to be really helping you. So, next we have Dr. Melnick and then 3 4 Dr. Farley. 5 DAVID MELNICK: I just wanted to go back to a point that Dennis raised. You know, in a 6 world where trial networks exist and there's a 7 8 capability for early initiative of strategy-type trials, is there anticipated flexibility about the 9 10 dissemination of that information? 11 You know, we need a mechanism for 12 communicating that strategic information to the 13 clinical community. And whether it's resistant pathogen trials or the application of a novel agent to 14 15 bacteremia or something else, what -- how do we get 16 that information out, you know, without incorporating that into the label? 17 18 MAN 2: (indiscernible) guidelines 19 again. CYNTHIA SEARS: Well, I don't know. 20 Ι 21 mean, guidelines is one approach but, Vance, you 2.2 presented several examples where strategic trials are

1	published in the literature and disseminated. You
2	know, it certainly would be helpful to clinicians to
3	have more concentrated information so you don't have
4	to go to individual facts. And also from the FDA
5	documents, it sounds like there's readily available
6	data early on that could be summarized while we wait
7	for publication. Sorry I didn't leap up. I was
8	thinking about it, but I welcome other input,
9	though.
10	DAN RUBIN: John?
11	JOHN FARLEY: Anything more on this
12	thread? Because I'm going to change the subject to
13	comparators. Anybody? So, Dr. Ucheo Uchea is
14	that correct? Dr. Uchea?
15	CHIBUZOR UCHEA: Yeah.
16	JOHN FARLEY: So, we really appreciate
17	you being here and appreciate Welcome's leadership in
18	really focusing on clinical trial capacity in the
19	areas where the pathogens of interest really are. So,
20	both sort of where you're headed as well as actually
21	some of the statistical methodologies which might
22	allow for smaller studies it feels like they're

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headed toward a -- using a comparator in those trials
 that may not be registered in those countries and may
 be expensive.

So, it may be early for you to ask this -- to ask this -- early to ask this question, but I wonder where you all are thinking about with that particular question.

8 CHIBUZOR UCHEA: Yeah, that is one of 9 the things that we've been thinking about. In terms 10 of comparators and standards of care, they vary quite 11 widely within -- within regions. And especially where 12 we're looking at as well, a big issue is availability. 13 And that's one of the key reasons why we're looking at the interoperability of our network with other 14 15 regional networks. And that will allow us to leverage 16 the expertise in those areas and also potentially 17 bring in the ability to use comparators that are known 18 and have -- that we have sufficient knowledge about in 19 other regions to be able to implement that into -into our potential region and have access to those 20 21 compounds for use.

22

DAN RUBIN: All right. I do have a

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follow-up to that issue. Just because of time, I
maybe want to give you a follow-up and then go to
Aaron, and then we'll go into the moderated
discussion. But some of the later questions are big
picture enough that I think they can address any
issues here. But do you have a follow-up?
MANOS PERROS: Yeah, it is about the
comparator but perhaps in the broader sense I'm not

8 comparator but perhaps in the broader sense. I'm not I'm trying to get my head around how this 9 a trialist. 10 If you -- if you enroll a patient with works. 11 pneumonia, at the point where you enroll him, you 12 don't know if it's carbapenem (indiscernible) or 13 carbon resistance, you don't know if it's enterobacteriaceae or pseudomonas. The standard of 14 15 care in the comparator, it's different for each of 16 those.

17 So, how does this work? A patient gets 18 admitted, you don't start them on Colistin, but you 19 would put them on Colistin the moment you diagnose 20 CRE, for instance, in some regions but not in others. How do we do that in the clinical trial network 21 2.2 context?

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1	DAN RUBIN: I don't have the answer to
2	that but, Aaron, did you have a comment? Or Dr.
3	Uchea, did you want to follow up?
4	CHIBUZOR UCHEA: Yeah, that is one of
5	the big problems that we know we'll be facing. It's
6	something that we really have to look at. And we very
7	much would value any additional input into that. But,
8	yeah, that's one of the key tests that we're going to
9	face.
10	AARON DANE: Yes. So, I just had a
11	question for Scott, actually. So, Scott, I really
12	like DOOR and what you can bring, but one of the
13	things in terms of designing a trial with that as the
14	primary endpoint is understanding the power and the
15	risk of the study.
16	So, imagine a company's running a study
17	and the company could sort of live or die by the
18	outcome of that study and how you (indiscernible)
19	and the assumptions you have to make. Because
20	obviously with all the different categories, you've
21	got to make a lot more assumptions than we normally
22	would do, which is normally just what, response rate.

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1	So, it's not using the approach, it's
2	more how to assess the risk of the approach in the
3	outset. I don't know if that's something you've
4	looked at.
5	SCOTT EVANS: Yeah. Well, there's two
6	ways to think about sizing the study depending on
7	whether you want to size it based on a DOOR
8	probability or a partial credit approach. And then,
9	you know, as with any other hypothesis test, if you're
10	going to do a hypothesis testing approach you've got
11	to come up with a null and alternative. And the
12	null's usually pretty obvious. But you've got to come
13	up with what sort of effect size you want to see. And
14	it's on a slightly different scale probability
15	you're better off in one treatment rather than
16	another, or and you've got to either take that at a
17	high level with just thinking at that probability
18	level or think about how there's going to be a shift
19	in the DOOR outcome between two treatments and try to
20	size to detect it.
21	Or partial credit is basically you're
22	on a continuous outcome. You can think of it as a

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as a difference in a difference in means or a
difference in proportions on a 100-point scale.
VANCE FOWLER: May I add to that? A
second element to respond to Aaron's very reasonable
question would ideally be to standardize the DOOR by

6 indication. And, in fact, sort of one of the early --7 sorry... So, the second strategy by which to avoid 8 some of the challenges that Aaron raised would be to standardize the DOOR endpoint such -- by disease type. 9

10 And, in fact, one of the goals of ARLG 11 2.0, one of the early initiatives that Helen, in fact, 12 is hopefully going to be leading, if I can persuade 13 her to, is to -- is to develop just that for the four most common anti-infective initiatives or approval 14 15 pathways. So, intraabdominal, complicated UTI, HAP/VAP, etc. That could then be publicly available 16 17 to all sponsors to utilize as -- probably an 18 exploratory endpoint. I doubt you're going to get to, 19 you know, a primary efficacy endpoint with a registrational setting. 20

But to make that available so that it's 21 2.2 a common standardized tool that could then be employed

1	with whatever anti-infective agent may be at question.
2	AARON DANE: Yeah, I think something
3	like that would be really helpful. And the reason I
4	raise it is because we often have a lot of angst, even
5	on what the response rate is. Yeah, and that's a very
6	simple measure. And we're not sure and that's
7	where a key risk can be.
8	So, if we're starting to move to three
9	or four categories, which clearly makes sense, is
10	clearly more sensitive, but then that means there's
11	much more consideration and much more risk with that.
12	So, anything that could help understand that in the
13	disease areas would be good, I think.
14	VANCE FOWLER: And I think it's very
15	likely that the factors, the conditions that impact
16	one patient's clinical response for an intraabdominal
17	infection is likely to be profoundly different from
18	that that he or she would encounter in the setting of
19	HAP/VAP or skin and soft tissue infection, what have
20	you.
21	So, the notion would be that you'd
22	create it carefully, validate it, and make it and

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include?

point?

Page 145 then essentially unleash it and make it, you know, publicly available for all sponsors to employ in their own development programs. JANE KNISELY: Do you have a related Okay, qo ahead. RIENK PYPSTRA: Yeah, just to continue on that, you're suggesting to develop these DOOR criteria for the purpose of clinicians. But as we discussed before, we have other people who are interested in this. In your talk, Scott, you mentioned patients. I think that is important. But we have in between category as well -- the payers, who are very important decision makers. So, if we include a DOOR outcome or analysis in our studies, which ones should we then Of course, we want to serve the clinicians but can we include another for the payers? Only to be concordant?

19 VANCE FOWLER: Yeah, sorry, I actually meant the intent and the intended audience of these 20 21 DOOR Tools would primarily not be the clinicians but, 2.2 rather, would be industry. Because the goal is to

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1	make available a tool much along the same lines as a
2	guidance. Think about the guidance for HAP/VAP or
3	ABSSSI or what have you. That this is a tool that
4	you know, Compound A, Compound B from Company A and
5	Company B, rather than create their own, have to go
6	through their own validation, etc., that this is a
7	standard external template against which to compare
8	the performance of that compound.
9	RIENK PYPSTRA: But the way to build it
10	
11	VANCE FOWLER: The audience is the
12	audience is indeed industry.
13	RIENK PYPSTRA: Yeah. But the way to
14	build it we heard from Scott that it was 43 profiles.
15	They submitted it to clinicians, they came to six
16	categories. I suppose if you submit those 43 profiles
17	to a payer, he may they may come up with a
18	different categorization. And so that's what I want
19	to bring in here, and it's actually a question to
20	Scott. Is that something that you've considered and
21	that is possible or what do you think about it?
22	SCOTT EVANS: Yeah. Well, the example

1	that I showed, we survey clinicians. But one of the
2	questions we get is well, maybe you should survey the
3	patients. Sometimes we forget about them in clinical
4	trials. I think so, two comments: One is I think
5	you could have multiple outcomes if in the sense
6	that if clinicians, or industry, or patients, or
7	payers are interested in different outcomes, you can
8	construct those different outcomes and analyze them in
9	studies.
10	One thing that we have done already
11	with the DOOR outcomes in various studies is when
12	we've implemented, for example, a partial credit
13	analysis approach, although pre-specification and
14	transparency is always the you know, so important
15	in clinical trials, when it comes to putting a value
16	on different outcomes, it seems to be avoided like the
17	plague.
18	And it's a very interesting dichotomy.
19	Transparency and pre-specification, but when it comes
20	to writing down a value or trying to value different
21	outcomes, nobody wants to touch it.
22	Now, acknowledging the fact that,

1 Number 1, that not everybody has the same value 2 system, including payers versus clinicians versus patients, or even within those categories, the 20-year 3 4 old woman is going to have a different value system 5 than the 70-year old man, and to acknowledge that. Now, one thing that we could do, as we 6 7 get our DOOR outcome, we say we want to implement a 8 partial credit approach. So, if I surveyed the clinicians and said, well, give me your grading key, 9 10 and as -- to adhere to transparency and pre-11 specification, I could survey experts in this field 12 and come up with sort of a population average of what 13 people feel a grading system should be. However, the actual analysis can also 14 15 portray that if you want to deviate from that value system and would like to do some other value system, I 16

17 can show you what the treatment effect is under that 18 value system. And we have examples of this. In the 19 CAZ-AVI Colistin study we did exactly this. And so, there are some mechanisms by which you can get 20 21 information about that value system. One is to survey 2.2 patients or clinicians or whatever the case may be.

The other is to get it in the course of the trial. 1 2 So, during the course of the trial, if you implemented, for example, a quality of life 3 4 instrument, and I set up a DOOR outcome. And if I 5 look at the patients in the most desirable category, those patients give me an idea using this quality of 6 7 life instrument -- they give me an idea of what their 8 quality of life is like. Then I look at the quality of life in 9 10 the patients in the second most desirable category and 11 that perhaps in using this instrument, that the ratio 12 of the quality of life of those patients compared to 13 the patients in the most desirable category gives me an idea about how to score it with information coming 14 15 directly from the patients themselves about how you impacted their life. 16 17 So, there's ways I can get patient 18 information, there's ways I can survey, if you want to 19 go to payers and other people, or go to clinicians who treat this disease, you can get that sort of 20 21 information. Don't be fooled that all of them are 2.2 going to have the same answer. And so the idea is, is

to collect that information, design and analyze trials 1 using that information, but acknowledge the fact that 2 3 people value things differently. That's life. That's 4 the way it is. 5 But I can show you how two treatments 6 contrast each other according to any particular value 7 system we like. And we have examples of this. 8 JANE KNISELY: Okay, so I have Rebecca over here has been waiting patiently, then Helen, then 9 10 Nick, then Roger, then Pam, then John. And we're all 11 on the DOOR question. 12 DAN RUBIN: And Sumathi told us to just 13 keep this discussion rolling since the next questions are kind of related. We're getting at those anyway. 14 15 JANE KNISELY: Great. Okav. Just to follow on the 16 REBECCA REINDEL: 17 DOOR tool, which I think is a really interesting way to look at benefit risk together. In the setting of a 18 19 known safety risk for a product where you use the

partial scoring system to assess the benefit and the 20

21

2.2 development of tools that are standardized for a given

risk of that specific safety concern -- with the

body site not allow you to be flexible in terms of
 assigning risk for specific safety concerns for a
 given product.

4 SCOTT EVANS: I'm not sure I completely 5 understand your question. But one point that came up 6 as I was trying to understand your question. When 7 you're identifying endpoints and constructing a DOOR -- so, the big concern in the room is if I take -- if I 8 try to construct some sort of ordinal outcome, could I 9 10 manipulate in such a way that I make my drug look 11 better than yours because of the way I construct it, 12 perhaps favoring some components of outcomes more than 13 others.

14 So, the outcome you're after is the 15 outcomes that are sort of important and valuable to They're not necessarily -- although, 16 patients. 17 obviously, you're evaluating toxicity associated with 18 particular interventions. But what you're measuring 19 is patient outcomes and things that are important to the patient. So, I want to make sure that when I 20 21 construct it, I'm not just including safety events 2.2 that happened to occur with this intervention, which

would -- but not this other intervention. 1 Because if 2 another intervention, your control intervention has 3 got their own sort of toxicity risk, if I either 4 emphasize those or deemphasize those, I'm not playing 5 a fair game. 6 So, what I want to measure is those 7 things that are important for patients or that we see that perhaps patients aren't aware of, but do it in a 8 fair way, not so specific to the intervention, 9 10 obviously there's some toxicity concerns you want to do with that. 11 12 But if you're going to be transparent 13 and open about how to construct it, of course, the concern of regulators and the concern of the audience 14 15 and people is that you don't build an outcome that is artifactually going to make you look better than the 16 17 next guy because of the way you do it. It's got to be 18 about meaningful evaluation of what's happening to the 19 patient. 20 HELEN BOUCHER: So, I just had a small 21 comment about the ability of DOOR, by route of Scott's 2.2 comment, to include aspects of the outcome that are

more safety related and for which there are values 1 2 that payers care about. And so, as opposed to the traditional non-inferiority trial where we have kind 3 4 of the standard safety display, DOOR could provide the ability to measure in clinical meaningful ways that 5 payers care about things like renal failure and things 6 for which there's not only mortality but also cost, 7 8 that people agree on that could help add to the value of a product, even if it doesn't add to the 9 10 indication, specifically.

11 NICK KARTSONIS: So, I've been thinking 12 about this door thing for many years now and I 13 actually find it very intriguing and actually think 14 it's a fascinating way to kind of think about marrying 15 up both the efficacy and safety aspects that kind of 16 go into the development.

And I do agree, I mean, trying to get the patient perspective and the payer perspective is important. But I will tell you with every country now having its own HTA agency trying to get agreement from payers is almost impossible. But, I mean, I do think we're also in a very unique situation that we've done

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1	all these clinical studies over the last decade and we
2	should learn from them.
3	And so I guess I had two questions, and
4	it's probably more, you know, directed to the FDA and
5	their perspectives on this. One is you guys have had
6	conversations with other regulatory agencies. Do you
7	think they would be open I'm trying to think about
8	this now, also trying to make Japan happy and Europe
9	happy would they be open to that?
10	And second is, is there a possibility
11	to use all these datasets that have been generated
12	that you guys have access to and do these sort of
13	analyses in a way that doesn't de-identify the
14	comparators, if you know what I'm getting at but
15	helps provide information that has value from an
16	endpoint standpoint?
17	So, I know it's a bit of a loaded
18	question but I'm just curious. Because I do think
19	it's an interesting way to think about things moving
20	forward.
21	SUMATHI NAMBIAR: So, for the first
22	question we haven't had a lot of discussion about DOOR

1 endpoint. At least -- more recently we've been 2 involved with one product. We've had a couple of 3 discussions. And in one of those discussions the EMN 4 was part of the meeting, so we met with the company 5 together.

I don't think we heard any sort of concerns in broad terms but it's more the details. And that's exactly what we were struggling with -- you know, what were the different components of the proposed endpoint? And so I think there is certainly a willingness to work. I just cannot speak for them and say they will agree.

13 But what I did not hear at the meeting was we just cannot move forward with this. But I 14 15 think for us, the bigger struggle was really defining which of the components were appropriate and how one 16 17 could actually discern because it was so -- again, I 18 don't have a lot of experience, but this one example 19 that we worked through, I think, between the different categories there was very little separation. And the 20 21 subjectivity that was involved in putting them in one 2.2 category or the other was certainly concerning.

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1	But there is room to improve and refine
2	that and we're working to do that. So I think Dan,
3	is that a fair assessment?
4	DAN RUBIN: I think that's fair. And
5	as far as using the existing databases to try to come
6	up with a better outcome, I mean, there's some
7	precedent for that through the FNAH project to look at
8	different datasets and try to define endpoint, but I
9	don't think there's been any discussions about using
10	datasets to try to inform more ordinal types of
11	scales.
12	SUMATHI NAMBIAR: So, as a follow-up,
13	Nick, so you meant not de-identifying the data
14	actually and that's where it gets a little
15	difficult, right?
16	NICK KARTSONIS: That's always the
17	tough issue, right? I mean, I'm speaking obviously
18	only for Merck but I imagine my counterparts might
19	feel similarly around it.
20	But on the flipside of it is it's
21	hypothesis generating, right? It's not hypothesis,
22	you know, proving. So, in some ways you could make a

1	case for helping to move the field forward. And
2	you've done that, right, with thinking about non-
3	inferiority margins and how you guys came up with M1s
4	and M2s and all that, which was incredibly helpful for
5	the field moving forward. So, I just raise that as
6	SUMATHI NAMBIAR: Good thought.
7	Certainly something we can talk about.
8	ROGER LEWIS: So, I think the
9	discussion about DOOR and the difficulties of, you
10	know, everybody agreeing on, for example, the relative
11	weights or even the ordering of different outcome
12	states is interesting. I think it's an example where
13	the more sophisticated approach to patient outcomes
14	illustrates problems with existing outcomes that we've
15	been able to not have to face.
16	So, for example, if you use 30-day all-
17	cause mortality, some people die in three days in the
18	ICU, some people die at day 29 in the ICU, some people
19	wish they were dead but they're still alive in the
20	ICU. And different people would value those outcomes
21	in a different order. But somehow 30-day all-cause
22	mortality just feels really solid. But it's the same

1 illusion. So, I think the discussion about DOOR 2 is really important and illustrative but I wouldn't 3 4 want it to be seen as a criticism of the approach. I think the approach is absolutely solid. The right 5 thing to do. And at some point you just have to --6 7 for the purpose of defining the endpoint of a trial, defining the statistical characteristics and what 8 9 defines an adequate and well-controlled trial, agree 10 on what the order of categories are by disease state, 11 just as Scott suggested. 12 But the fact that we don't have those 13 same level of discussions for the rich outcomes that occur in the people who do or do not die in 30 days, 14 15 that's just because we're choosing not to have those 16 discussions, not because the same considerations don't 17 exist. 18 JANE KNISELY: I think I have Pam next. 19 PAMELA TENAERTS: So, I think this is a fascinating way to look at a novel endpoint and I 20 21 really think there might be a need for something to 2.2 supplement the mortality, which is sort of

1	unidimensional in many ways.
2	But what I would say is there's
3	opportunity for confusion if you do DOOR as per the
4	physicians, DOOR as per the patients, DOOR as per the
5	payers, and then you're kind of like, well, which ones
б	weigh more in the decision?
7	So, my thing would be well, combine
8	them all and come to a consensus where everybody it
9	might be lower but everybody can agree to it. And I
10	think that might be, you know, baby steps instead of
11	sort of setting a new thing and potentially creating
12	additional confusion.
13	VANCE FOWLER: Yeah, agree. And, in
14	fact, one of the sort of work streams that's intended
15	is obtaining the essentially, a patient-assessed
16	quality of life assessment for those same four
17	indications. We've already done it with staph aureus
18	bacteremia and gram negative bacteremia where we
19	establish quality of life in those patients.
20	And what was striking was two things.
21	I guess the clinicians in the room would probably
22	guess that the impact my hypothesis was that staph

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lives.

Meeting Page 160 aureus is going to -- it doesn't tend to kill you right away, it just sort of chews on you a long time. And that's exactly what you found -- that these patients who had ostensibly had been evaluated as cures, when you interviewed them extensively, you know, six weeks later or what have you, they were talking about, you know, terrible impacts on their Having to wear Depends because they can't get to the bathroom in time. I mean, a profound impact on the quality of life. And these were ostensibly cures, what would've been cures. So, and for gram negatives, it was really all about the first two weeks. If they're done So, yeah, incorporating that element,

in two weeks, by six weeks their quality of life is 14 15 much higher than that with staph aureus. 16 17 that's, again, part of the intent of creating this 18 DOOR endpoint, is building patient assessment into 19 that step forward. JANE KNISELY: John Rex? 20 21 JOHN REX: So, the build I've got 2.2 around DOOR here is that there -- I hear the plea

about people don't want to rate the number -- they 1 2 don't want to put a number on it. There actually has been work done that I think maybe we just need to 3 4 leverage it better. And I'm looking right now at the University of York and the University of Sheffield 5 report that the U.K. is using to construct its value 6 7 arguments around conducting a pilot program of buying 8 new antibiotics. And there's a whole bunch of stuff in here about health-related quality of life and how 9 10 you might think about measuring that in the setting of 11 any infections. 12 And I've not decoded all of it and 13 thought it through, but it actually makes me think further back to the ERG report from three or four 14 15 years ago where I first learned about the value of a 16 statistical life here and the way that the actuarial 17 community estimates the value of staying alive for a 18 period of time, and it feeds into the quality adjusted 19 life here concept. 20 So, there are some tools that exist 21 that would be familiar to the health technology 2.2 assessment world that we may not know about. And so

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my thought here about D	OOR is, it's very i	nteresting -
- and that plot about t	he vancomycin showi	ng that the
lower exposures were ac	tually net better,	at least
from my perspective, ex	traordinary.	
So, this	notion that the ID	community
needs to start folding	in some of those me	asures
and I'm just compliant	that some of it exi	sts and
that's the theme here.	We need to borrow	that
information and apply i	t here.	
AARON DA	NE: I was just goi	ng to say in
relation to that as wel	l, the other area w	here there's
a lot of work done with	quantitative benef	it-risk
assessment, which I thi	nk we should look a	t as well
for the same reasons.	Because that's all	about being
very clear on what the	trade-offs are, and	you're very
explicit about how you	weight them and you	can look at
what happens if you cha	nge the weighting.	So, I

what happens if you change the weighting. SO, I think, again, that's something which is similar which we can draw on.

SCOTT EVANS: Yeah, just one comment to Thank you for the information about the quality that. of life. DOOR has been used in other disease areas,

1 including stroke prevention trials, for example, where 2 there has been a lot of quality of life valuation 3 about what happens to patients -- strokes, MIs, and so 4 forth, where primary endpoints trials are time to 5 stroke, MI, or death, or maybe something else.

6 But deaths are worse than strokes,, and 7 strokes with permanent consequences are worse than 8 things that are transient, and you can factor in 9 bleeds and all that sort of stuff. And there has been 10 scoring systems set up in that area and/or analyses 11 that have used that information.

12 So, I wanted to address a AMY LEITMAN: 13 few points that were discussed just from the perspective of somebody who's sort of getting 14 15 entrenched in the world of PROs and PRO development because it's difficult to get patient feedback and 16 17 looking at the different kinds of infections. 18 Survival is not enough for a patient. That's really 19 the goal, right, when you're treating a patient? Generally speaking, you do want them to survive. 20 21 And I heard what was said before, that, 2.2 you know, what a patient wants when they're 20 and

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what a patient wants when they're 70 or 80 or 90 may 1 2 be different. And you may get to a point where a patient decides that they are not willing to continue 3 4 on a journey. But, generally speaking, the goal is for the patient to live, to survive the treatment and 5 survive the infection. But you're not treating an 6 7 infection, you're treating a patient. So, how they 8 survive matters.

And I think what we see when we are 9 10 dealing with patients, a lot of times what we see, and 11 maybe this isn't seen as much with the ID physicians 12 because the follow-up is done by another physician 13 probably. And maybe it should be done more by an ID physician so they could assess how the patient is 14 15 doing. And it might be an easier way to catch if 16 there's going to be a reinfection, there might be 17 earlier signs that a physician could catch.

But you have to look at how that patient is doing. And so often we see them and it's just -- they still feel like a train wreck. And it's months and months later -- and this is with all kinds of respiratory infections, and then you have to deal

1	with skin and soft tissue infections. If there's been
2	a surgical debridement, don't assume that there's no
3	permanent impact. They're worried about disfigurement
4	and they have to live with that impact.
5	And, you know, we actually deal with
6	that with NTM. We catch the patients who are
7	disseminated NTM and, unfortunately, they've gone to
8	those, you know, cosmetic surgery centers and gotten
9	those infections and I mean, there were I think 17
10	deaths in New York City alone last year from that.
11	And that's, I'm sure, not the only pathogen that's
12	causing these infections.
13	So, you really you know, even if
14	it's like an abscess on a limb, if it gets really bad,
15	at some point you start thinking they start
16	thinking am I going to lose my limb? Like, what's
17	going to happen? So, there's a lot going on.
18	And then after that, let's say it's
19	something with an abscess, you clear the abscess, it
20	
	heals. Every little thing they become paranoid about.
21	heals. Every little thing they become paranoid about. And it's something they get a bug bite and it's,

1	again?	What	am I	going	to	do?	Are	they	going	to	be
2	able to	treat	it?								

3 So, it's really important to understand 4 the patient journey, not just at the start of it, not just while they're going through the critical event, 5 but afterwards. And really not just a month 6 afterwards. For some of these patients it continues 7 8 months or even years afterwards. Their body has undergone really a brutal assault. Some of these 9 10 medications are really hard to take. Their immune 11 systems have completely gone haywire trying to fight 12 this infection. They're exhausted. They probably are 13 having some kind of a nutritional deficiency, because when they're fighting an infection, they're just 14 15 burning through calories.

You know, one of the things that our organization is starting to develop is a pamphlet on mental healthcare and a pamphlet on nutrition care for infection because those are things that patients have asked us for. And when I've talked to -- you know, I've talked to a sepsis survivor who was like, you know, the nutrition thing would probably be helpful

1 for people like us, who, when we're coming out of 2 this, they've lost so much weight and they need 3 guidance on not just how to put on weight but how to 4 do it healthfully.

5 Like, you can't just pile in Ben & 6 Jerry's as much as we'd like to. They can't just keep 7 eating Ben & Jerry's all day. They have to figure out 8 how to balance... You know, yes, you can have some of 9 that but you have to balance the good nutritional 10 intake and put those calories back on, and give your 11 body what it needs to heal as well.

12 There was one other thing I wanted to 13 talk about with respect to payers, and this is something that I'm seeing a lot in the patient 14 15 advocacy community. So, the Institute for Clinical 16 Effectiveness & Review, ICER, has a standard that they call Quality-Adjusted Life Years, QALY, or "Qually" as 17 18 we call it. They start looking at various treatments 19 and whether those treatment costs are justified by quality-adjusted life years. 20

21 And one of the things that we're 22 dealing with now is that antibiotics are not

compensated well enough. And we're looking at
 developing these novel treatments, and it's really
 expensive. And these treatments are going to be
 expensive, and it's justifiable considering the cost
 and considering the value to society.

But we're talking about some of these 6 7 infections like they are rare diseases. One of my 8 concerns is that we'd better start bringing payers into the fold on this conversation as well as patients 9 10 because I can guarantee you at some point down the 11 road, ICER will turn its attention to some of these 12 treatments and say, well, but, you know, two years 13 later, they're still -- they're still really sick or they're getting sick again, they're getting pneumonia 14 15 That is not -- I mean, believe me when I tell aqain. you that patient advocacy groups will kick up a huge 16 17 fuss because any time this happens, it just seems like 18 a particular group ends up getting targeted.

19 The patients suffer most. It's really 20 up to the patient to decide what the quality of life 21 is that they want afterwards, you know? Is the 22 quality of life that they're going to get from the

1	treatment acceptable?
2	And I just wanted to you know,
3	Roger's point about 30-day all-cause mortality really
4	goes back to measuring out over time with PRO.
5	Measuring out 30 days isn't enough. Measuring out 60
6	days might not be enough90 days. So, it's we're
7	at that point now where we have to start asking these
8	patients what is your journey like? How long is it
9	taking? Because when we start incorporating patient
10	input into these clinical trials and we start
11	measuring with PROs, one of the most important
12	questions and this is the one that we're grappling
13	with now, it's one of the many how long do we
14	measure out?
15	And it's critical because it can tip
16	either way when you're trying to decide do we approve
17	this drug or not? Because you're looking at it
18	well, is there clinical benefit? You may not see the
19	full impact on the patient for 12 months or longer.
20	It may take them that long to recover from any kind of
21	infection state if it's that serious. So, we need to
22	know for sure like, is there an extended period of

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1	time that we measure out? So, we need to start
2	finding these patients and talking to them, and
3	listening to them.
4	JANE KNISELY: Thanks. I have Sara and
5	maybe well, you're making your comment. Ah, thank
6	you.
7	SARA COSGROVE: So, I wanted to circle
8	back to Rebecca's comment about toxicity real quick
9	with DOOR. Because the two big studies that have been
10	done with DOOR related to toxicity have been done with
11	Vanco and with Colistin, for which we have the luxury
12	of 60 years of knowledge that both of those drugs are
13	nephrotoxic.
14	And I think unless you have a signal
15	already from a Phase I or II study, that if you're
16	trying to do a DOOR analysis along with the
17	registrational trial, you may be missing significant
18	toxicity issues when you construct the DOOR. And I
19	don't know if you've thought about those issues or how
20	to handle it, or maybe that's a flaw that's tough to
21	deal with and maybe DOOR is better done after, you
22	know, the registrational trial.

1	VANCE FOWLER: Yeah, I totally agree.
2	That's a great point. And it's one of the reasons why
3	I feel like the ultimate contribution of DOOR, at
4	least in the short term, will be in exploratory or
5	secondary endpoints. I think the primary again,
6	with, you know, the responsibilities that the FDA is
7	charged with in terms of evaluating safety and
8	efficacy I don't see a way around the safety
9	component not being primary and not being evaluated
10	independently for precisely the reasons you brought
11	out.
12	So, I'm suggesting that DOOR Has a
13	meaningful contribution for drugs that are already
13 14	meaningful contribution for drugs that are already available and trials that are strategy trials, we
14	available and trials that are strategy trials, we
14 15	available and trials that are strategy trials, we can consider it as a primary. But I think for the
14 15 16	available and trials that are strategy trials, we can consider it as a primary. But I think for the purposes of evaluating new compounds, safety is going
14 15 16 17	available and trials that are strategy trials, we can consider it as a primary. But I think for the purposes of evaluating new compounds, safety is going to have to be paramount. And, Helen, did you want to
14 15 16 17 18	available and trials that are strategy trials, we can consider it as a primary. But I think for the purposes of evaluating new compounds, safety is going to have to be paramount. And, Helen, did you want to step on that?
14 15 16 17 18 19	available and trials that are strategy trials, we can consider it as a primary. But I think for the purposes of evaluating new compounds, safety is going to have to be paramount. And, Helen, did you want to step on that? HELEN BOUCHER: I would just add that
14 15 16 17 18 19 20	available and trials that are strategy trials, we can consider it as a primary. But I think for the purposes of evaluating new compounds, safety is going to have to be paramount. And, Helen, did you want to step on that? HELEN BOUCHER: I would just add that by the time gets to Phase III, we usually know

1	kind of toxicity we know we might be facing with this
2	drug or class of drugs. Then there's that which we
3	don't know, which Scott was talking about. And I
4	think the DOOR, if you're going to apply it in a Phase
5	III setting has to think about both. But the fact
6	that you can do that is actually more than we can do
7	in some standard trials. So, that's where there might
8	be a benefit.
9	SCOTT EVANS: Maybe if I could just
10	follow up. I think one thing a couple of
11	points. One is if you use a DOOR endpoint you should
12	analyze all the components to it. In and of
13	themselves, they may or may not be interpretable. As
14	I mentioned, you know, you measure duration of
15	hospitalization, whether you can interpret that, you
16	know, it depends on what's happening with other
17	things.
18	I do think, though, that when you do
19	get to late phase trials you will have an idea about
20	toxicity. That's what the early phase trials are for.
21	However, I do think that in the long run, the vision
22	for a DOOR outcome, that the safety components that

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1	weigh into it should not necessarily be intervention-
2	specific.
3	So, if you're comparing two treatments,
4	one may have a toxicity in one place and another has a
5	toxicity in another place, the importance of a
6	relative importance of those two things depends on
7	their impact on patients. And so in some ways you
8	want to be agnostic to sort of intervention-
9	specificity. What you want is an outcome that
10	evaluates the impact on patient lives.
11	One frustration I've always had in
12	trials is we have all these rating systems for AEs
13	severity, seriousness, is it related to treatment? Is
14	it not related to treatment? Treatment-emergent, etc.
15	What I would like to see is a rating that says how
16	impactful is that adverse event on the patient? And
17	that's actually what you would like to factor in.
18	JANE KNISELY: Sue, did you have a
19	related point?
20	SUE CAMMARATA: No. Actually, I wanted
21	to talk about clinical trial networks.
22	JANE KNISELY: Okay.

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Page 174 SUE CAMMARATA: But I didn't know what the time you'd want, because according to the agenda, it looks like this is wrapping up at some point here soon. JANE KNISELY: Yeah. We are well into our moderated panel discussion, I think. So, let's just go ahead and note the discussion questions. So, we talked about a lot of things this morning. DOOR was a very interesting part of that but not the only part. And so I think we should open this up now to additional discussion. So, please go ahead. So, for me, I wanted to SUE CAMMARATA: talk a little bit about the clinical trial network. Because I think on paper that always sounds fantastic in some ways. And I was actually part of the Pew meeting -- if it was out of 2016 or 2017, it just -time flies. And I recall at that meeting, and I think it still is an issue, is what is the design and intent of that trial network? If I recall, most of the pharma representatives were quite concerned about things like -- I'll say master protocols and that because of the

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Page 175 challenges of a particular drug, particular comparators, time, you know, how would you use this data? I would say that I'm particularly happy to see that -- at least the Pew Trust is talking about pediatrics. I've said at that meeting and I still say it's not sexy, it's not exciting. We've talked about pediatrics. But if you want to be successful, I truly think that that is the one place3 you can start to get pharma potentially involved because it's less

11 competitive, there' less risk. It's not really a
12 money-making area for most companies. But it's one
13 that every single company, whether you're big or
14 small, has to...

15 I mean, I'm currently responsible for three pediatric programs, you know. 16 So, it's 17 something you have to spend money -- and it might be 18 an area of the most success, but it absolutely is not 19 sexy to everybody else except the pediatricians in the But it is one area that I would highly 20 room. 21 recommend thinking about because -- for any of the 2.2 trial networks.

1	But I am concerned and I'm curious in
2	this meeting, compared to that previous meeting, about
3	the thoughts of everybody else about master protocols.
4	I mean, for me, a clinical trial network that might be
5	able to be up and running, cost effective, just to get
6	a contract signed with an academic institution in the
7	United States can take six to nine months. And that
8	is always one of the other challenges of being able to
9	do trials in the U.S. So, having that available with a
10	site that's motivated is interesting. The problem is
11	I'm not sure there's enough volume over time to do
12	some of these trials.
13	So, for me, I was interested in this
14	concept of sort of the idea of the strategic trial
14 15	
	concept of sort of the idea of the strategic trial
15	concept of sort of the idea of the strategic trial that might not be pharma-sponsored but might be
15 16	concept of sort of the idea of the strategic trial that might not be pharma-sponsored but might be sponsored in some other way to get information to keep
15 16 17	concept of sort of the idea of the strategic trial that might not be pharma-sponsored but might be sponsored in some other way to get information to keep some of these sites open. For example, in HAP/VAP, to
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15 16 17 18 19 20	concept of sort of the idea of the strategic trial that might not be pharma-sponsored but might be sponsored in some other way to get information to keep some of these sites open. For example, in HAP/VAP, to answer interesting clinical questions that clinicians want to have answered but then also could be a site that participates in a clinical trial since for

might want to have something different because of the
 nuances of their product.

So, I'm just curious about h ow other pharma companies feel about it because I see lots of discussion but I'm not sure about the reality, for example, of a master protocol type system being set up versus something that's more flexible to handle both those strategic kind of questions and then the registration type of questions.

10 JANE KNISELY: Thanks for bringing that 11 up, Sue, because I had the same question. I also was 12 at that 2016 meeting and heard the same thing you did. 13 So, we heard a little bit of it yesterday from David, so I'm curious to get some perspectives from companies 14 15 about what are their perspectives on these clinical trials networks? How could they be useful? Sort of 16 17 this third question here. So, go ahead, Nick.

18 NICK KARTSONIS: So, thanks, Sue, for 19 raising that. In fact, the one area of it we've 20 talked about internally where we'd love a network was 21 pediatrics. I mean, because I do think you've raised 22 all the right issues around that, that I think would

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1	have value in terms of that.
2	You know, I sort of harken back to our
3	recent experience with HAP/VAP. And each of our three
4	studies was different. And we actually talked about
5	this. Could we have done this in the network
6	situation? But they each looked at different
7	endpoints and they had different visits to some
8	extent, and just some of the inclusion-exclusion
9	criteria also varied a little bit.
10	You know, there's a host of things we
11	worry about around the network. You mentioned master
12	protocols and the factors that go with that, but then
13	there's all the operational stuff of different safety
14	tests for different drugs, how do you handle that?
15	Blinding, comparators, how do you handle all of that
16	stuff? And then there's the inevitable database
17	issue, which, every company has its own database. And
18	I don't know how other companies are but every time I
19	feel like we've run a database, the lights dim a
20	little bit at Merck, then they come back on. (Laughs)
21	And so it isn't without its shares of
22	challenges of trying to share that data across the

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1	databases. So, just mention that.
2	JANE KNISELY: Maybe we can hear from
3	Rienk and David, and then go to Chibuzor.
4	RIENK PYPSTRA: Yes. I also agree that
5	pediatrics is a very obvious one. I think HAP/VAP is
6	also a possibility. It all depends on the level of
7	integration. If you have a network of independent
8	sites, you can indeed run a common protocol and you
9	can organize yourself, and you may have some savings
10	on the patient numbers and therefore on cost.
11	But you could go with the integration
12	all the way to having a common database for the whole
13	network. And the patient data that you're collecting
14	in a continuous fashion are immediately available in
15	that database. And then the step of adding an
16	investigational or an interventional arm in that
17	becomes much more simple.
18	It also overcomes some of the issues
19	that we had about informed consent that were discussed
20	this morning. If you were such a site, every patient
21	admitted to your site would enter the hospital or
22	enter the ICU. You can determine yourself where you

1 put that barrier. And at admission, get a 2 questionnaire. We are a research site. Would you 3 agree to participate in research? 4 If the patient says yes, from that 5 moment on, the data of that patient go into the

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database. Whether you will or not need that patient
doesn't matter yet. And then when you have a specific
protocol, you re-consent the patient. You are now
qualifying for this intervention. Do you still agree
to participate in the research? That would be
something that would facilitate it.

And such a network would have quite a lot of advantages. First of all, it would be of a very high-quality standard. The data that you're collecting in each site would be comparable because that's how the database has been set up. You don't have different physicians collecting different types of information in their patient loads.

So, I think that would be a very important homogenization of the data and making sure that we can compare the data across sites.

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DAVID MELNICK: So, I agree with much

1 of what's been said. I think -- you know, perhaps we 2 could uncouple the idea of a trial network from a 3 master protocol. We don't have to do the whole 4 shebang at once, you know. And there is the obvious 5 saving -- you know, I love this term of a single use 6 network.

7 We reinvent the wheel with every one of 8 these trials. This has confused me from my start in industry, you know, 25 years ago. What do you do when 9 10 you start a trial? You look at the FDA guidance and 11 at the three last trials that had been done and, you 12 know, you write your protocol and then take it to the 13 agency and get it approved, and you're off assembling a trial network. 14

15 But those frontend activities -negotiating the clinical trial agreements, you know, 16 17 it's an incredibly laborious task and we pay an arm and a leg to CROs to do that work for us. I mean, it 18 19 would seem to me that having a shared infrastructure that could include things like shared SOPs, you know, 20 21 common CTAs, and then gradually move in the direction 2.2 of, well, if there's a HAP/VAP network, we work on

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1	endpoints. They're going to work and gradually move
2	toward a master protocol that would be acceptable.
3	So, it makes sense to me.
4	CHIBUZOR UCHEA: I totally agree with
5	you, David. The two concepts are independent
б	they're complementary but independent. We really need
7	to be focusing on the initial development of the
8	network, and that's what we've prioritized. The use
9	of continuous master protocol is a brilliant idea.
10	It's one that requires it's much more resource-
11	intensive. We're a bit of a way from that at the
12	moment. It requires further development and
13	assessment of the investability.
14	Going back to clinical trial network in
15	general, the one that we're setting up in Southeast
16	Asia, we really want it to be able to facilitate
17	parallel follow-on studies but the smaller type and
18	optimization studies. And that's one of the reasons
19	that I'm really happy that people are talking about
20	pediatric indications, because that's something that
21	we see as a key area that we can achieve a lot of buy-
22	in. We want to get away from the whole waiting ten

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1	years for a product to go through and start studies in
2	pediatric indications. And we're working with Penta
3	to try and leverage their expertise in this area and
4	be able to build up our network to be able to run
5	studies like that.
6	And I'll also go to a point that Vance
7	made earlier, which is really important, about
8	funding. We really need to be able to try to leverage
9	government funding for these kinds of projects.
10	They're heavily dependent on philanthropic
11	organizations. But the resources needed, they're
12	really deep.
13	Another way we're setting up our
14	network is also for our investigators to be able to go
15	out and get their own ground funding as well. So, our
16	secretariat function will be able to support them in
17	proposal development.
18	And I'd like to go back to a point that
19	was made yesterday about bureaucracy, the potential
20	for this to be like a heavily bureaucratic way of
21	operating. One of the key reasons we're operating
22	with a network secretariat is it will effectively work

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1	as the external business face of the network. So, it
2	will provide a single point of entry.
3	So, the whole idea is moving away from
4	that bureaucratic approach where you're constantly
5	trying to find various different trial sites. It's
6	one point of entry, and then we build from there.
7	JANE KNISELY: Okay. So, a follow-up
8	point on this topic? Sue? Okay, go ahead.
9	SUE CAMMARATA: Well, one of the
10	questions I did have for you around the pediatrics
11	and this is somewhat off topic but I'm interested to
12	ask is around using those sites for drug
13	development. Because, typically, you try to avoid
14	going to places where you're not going to
15	commercialize. It's just the ethics of it all. And -
16	- but, unfortunately, with resistant bugs you have to
17	go where you can go.
18	All the companies would love to be
19	commercialized globally but it's not necessarily
20	happening. So, is there some arm of that involved in
21	this discussion that the Pew is taking about for
22	products that are eventually studied? And then some

way to have access -- especially for the small 1 2 companies that don't have partners in those areas. 3 CHIBUZOR UCHEA: In general, access is 4 something that's really important to our organization and it is something that we're looking at. The key 5 for us with selecting Southeast Asia as our anchor 6 7 point is the access to public patient populations with 8 hyper (indiscernible) disease and being able to look at the multidrug resistant and extensively drug 9 10 resistant indications. 11 It was really refreshing to hear 12 Steven's talk yesterday to see that this is being 13 further explored, and that the data presented less concerns around generalized ability because that is 14 15 one of the key reasons why not a lot of clinical 16 research is done in these areas. 17 And what we're trying to do is we're 18 trying to leverage that, we're trying to build the 19 capacity, improve the trial site so that regulatory bodies are happy with the data. The big thing is 20 21 about the value of ex-U.S. and ex-EU data. So, 2.2 initially operating this trial site in this region

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1	will help us to be able to leverage that and to show
2	the potential advantages of being able to run clinical
3	trials in these regions.
4	SUE CAMMARATA: This is actually a
5	follow-up but it's directed more to the U.S.
6	clinicians, because I keep hearing about the concern
7	about not U.S. data. And I think for most of the
8	pharma folks, and FDA, and other agencies, they're all
9	comfortable with there are differences but this
10	data is translatable.
11	And so I'm just curious because I know
12	that there is this concern about the lack of U.S.
13	patients. But I did a trial recently where we went
14	out of our way and spent extra time and money to try
15	to enroll in the U.S. and we got five patients versus
16	860 outside the United States. And that was a
17	significant effort. We actually delayed our timelines
18	to try to get those U.S. patients because everybody
19	talks about it.
20	So, I'm just curious since many of the
21	people here actually accept that. But I keep hearing
22	the U.S. talk about it and I'm just questioning

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	concerned or questions about how you feel about that.
	Because that is where I think, for the bugs, I mean,
	that's where we're going to have to go.
	VANCE FOWLER: Yeah, I mean, I'd put
	that on the scale of concern about right up there
	with like a meteorite hitting the planet. It's just
	not it's just not a real concern. I'd much rather
	have, you know, data than not data. And, okay,
	there's going to be practice variation. I mean, we're
	setting up sites and got sites going on in China and
	all these other places. It's definitely different.
	But guess what? You can randomize
	folks there. You can block randomize. You know,
	there's means by which you can address that. And at
	the end of the day, if you're you know so, no,
	it's not to me, it's not a meaningful concern.
	JANE KNISELY: Additional responses?
	SUE CAMMARATA: I was just going to say
	the same thing. I don't think I actually don't
	even hear this as a concern brought up amongst other
	infectious disease doctors. I think people would much

22 rather have data in patients with resistant organisms.

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1	HELEN BOUCHER: So, I will agree with
2	my esteemed colleagues, but also just raise the point
3	of view that we're in America, the most resource-rich
4	country, and the question for us is, is it acceptable
5	that our patients are not in these trials? And the
б	questions are being raised by payers, and journal
7	reviewers, and editors, and others, when there are
8	zero or five out of 800 patients, there are some
9	people think there are ethical issues.
10	So, I would just submit that it's at
11	least worth of our consideration before we say, well,
12	we're just going to give up on this enterprise in
13	America.
14	VANCE FOWLER: I'm pretty read to give
15	up. I mean, I think I'd much rather I'd much
16	rather go ahead and get actually try to get some
17	trials done and get some answers and some data
18	somewhere. I know, I get banged around about the
19	journalizability and all the ex-U.S., yadda-yadda, but
20	for crying out loud, if we weren't in a crises, we
21	wouldn't all be sitting here and investing two days of
22	our time on this issue. And that's from all sides. I

1 think -- we're in a crisis, folks. The building's on 2 fire, you know, and we need to do something. We need 3 to get some data, we need to understand how to treat 4 patients, do the right thing. Crawl, walk, run, fly. 5 Thank you.

JANE KNISELY: John looks like he has a burning response. Pam's had her card up forever, so we'll go to her after John. And then I think we do want to give the audience an opportunity to ask some questions. So, if you have one, please come to the microphone and we'll continue to work here in the meantime. So, go ahead, John.

JOHN REX: So, Helen, are you concerned that we're in a position of looking like we're using the rest of the world as our guinea pig? I mean, is that sort of the idea you're getting at, that we need to be sensitive to that notion?

HELEN BOUCHER: I think that's part of the issue, and I think it's also the question of whether our patients deserve the chance to be included in these trials. And what we're seeing is that they're not. 1

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Page 190 So, I totally hear Vance and Sara and the need to get on with it, and we cannot let the perfect be the enemy of the good. LINDSEY BADEN: But there are multiple problems to be solved here. I mean, we need high quality data, full stop. I agree with Helen. It's a -- it's a shame that that can't also occur in the Americas. But I don't think they're exclusive. On the other hand, we need high quality data more quickly. JANE KNISELY: Okay, Pam. Sorry. PAMELA TENAERTS: I'm actually not a patient person but... That's actually the point I wanted to make. So, with City, we started this effort in antibacterial work. A lot of the comments I got from people in the field was, well, we've been talking about this for ten years. What is going to be different about this time? May I remind you that that was in 2012? And it feels like we're still having the same discussions over here. So, what I would like to say is, you know, we are underestimating the risk of what

1	we're doing right now and overestimating the risk of
2	novel approaches. And I really would like you guys
3	and girls because I've been told that you can't
4	really say "you guys" because it's discriminatory
5	that all of you sort of have an open mind. And when
6	people have new ideas, to test those ideas. Make
7	funds available. And I'm looking at the NIH because I
8	don't know where else the money's going to come from.
9	But to test those opportunities for new ideas.
10	Because to just go off and do new things new isn't
11	always better. Because if we thought everything was
12	going to work then why are we even doing the clinical
13	trials to begin with? Because everybody is convinced
14	their drug is going to work when they go into
15	development, right? Otherwise you wouldn't do it.
16	But we've been proven wrong.
17	No reason to think that some of these
18	things that we think are going to work to make this
19	better may be wrong too and have unintended
20	consequences. But I would like to say, like, just
21	keep an open mind and start doing something. Maybe
22	just start, and that'll change things.

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1	I mean, you guys have come a long way,
2	because when I talked about master protocols in maybe
3	2014, I heard, oh, that doesn't work in our field.
4	So, I'm not saying but there's other things. Beg,
5	borrow, and steal from other specialties. I mean, you
6	know I don't know, that's just what I would like to
7	say.
8	I think what we find the most difficult
9	with the work we're doing is coming up with
10	recommendations as people, you know, go back to the
11	status quo, it's easier to sort of do what you're
12	always doing than do something new. So, anyway
13	JANE KNISELY: Roger, you've also had
14	yours up for a while.
15	ROGER LEWIS: So, I wanted to make two
16	comments about the issue of out of U.S. data and
17	applied to U.S. patients. The first is just a request
18	for statistical clarity when we're discussing the
19	differences. There is a tendency, at least in
20	informal conversation, to confuse differences in the
21	overall average outcome of patients in different
22	locations from the expectation of the treatment of

1	fact.
2	And the thing that is scary in terms of
3	the use of out of U.S. data to inform U.S. regulatory
4	decision making is if you think the treatment effect
5	is heterogeneous as opposed to the background success
6	rate of disease. And I just urge us when we discuss
7	this, regardless of the position we're taking on it,
8	that we're very clear to make a distinction between
9	differences in prognosis and differences in efficacy.
10	The second point I would make is that -
11	- and I think Helen was also referring to the Belmont
12	Report principle of justice, that the population in
13	whom the risks of research are borne should be the
14	population that benefits from the results of that
15	research. I didn't say it very well but it's in
16	print.
17	That can ideally, there's a one-to-
18	one correspondence between the population on which the
19	experiment is conducted and those who benefit. But
20	you can when that is infeasible, you can partially
21	address the inequity by making sure that the benefit
22	of the research is made available to the population in

1	which the research is conducted.
2	So, I don't want to I'm not in a
3	position to comment on the feasibility of doing these
4	studies in the U.S., and if doing so delays the
5	availability of the agents globally, that's an
6	important negative aspect for everybody. But we
7	should address the question of whether these agents
8	that are being developed are then available in the
9	countries in which the research is conducted.
10	RYAN CIRZ: I'll be brief because that
11	was essentially what my point was, was hearing
12	Well, I guess, first worrying about ex-U.S. predicting
13	U.S. I mean, we do use mice to set our dose initially,
14	if that makes you more comfortable. And, generally,
15	if you are really rigorous, you get it right every
16	time. We don't see a lot of failures except for sort
17	of dosing errors that a lot of people think could've
18	been predicted.
19	But that is a worry, and I think about
20	this a lot. And regardless of whatever you thought
21	about the plazomicin study, we did most of it in
22	Greece and the drug has not been back to Greece in

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1	years. And whether you believed it had that effect or
2	not, those patients don't have access to it. And that
3	is what's going to happen. And there is that little
4	bit of risk of saying, let's go study it over there
5	where we don't have a lot of it. Let's go study it
6	over there and then we'll bring it back here to
7	protect ourselves. And I think someday we've got to
8	be prepared for how to handle that properly.
9	DAN RUBIN: All right, so we have
10	another online question. Roger, this is a follow-up,
11	I believe, from the earlier John Tamico question about
12	borrowing. And remember, the earlier question was
13	about CIAI, and the follow-up is that "The issue is
14	not just a difference in treatment effects but in
15	noise from source control and anticipated small
16	numbers of patients at each body site. So, it is a
17	body site issue, not an antibiotic site issue. Should
18	this disqualify CIAI?"
19	ROGER LEWIS: Great, thank you. So,
20	the hierarchical model takes into account the signal
21	to noise ratio within each of the sites or the
22	subgroups. The signal to noise ratio is influenced

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both by, obviously, the number of observations you
 have within that group and also the variability you
 see.

4 So, if there's a source of nonbiased 5 noise, just randomness in the adequacy of surgical site control or ancillary therapy, that can be handled 6 pretty well within the model. If what's happening is 7 8 the source of noise systematically makes a drug ineffective or, I guess, conversely, more effective --9 10 but ineffective in a site, then that's an issue 11 because then the site really isn't exchangeable. You 12 know something's different about that site than the 13 other sites.

And, statistically, it's completely analogous to a setting in which you know the drug doesn't get there. If you knew the drug doesn't penetrate the meninges even when they're inflamed, it shouldn't include meningitis as one of the infection types in a hierarchical model because you know it's different.

21 So, this is -- there was a brief 22 comment in my presentation where I said that the

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1	decision to include a site, or it could be a disease
2	type, in a hierarchical model should be based on pre -
3	- in learn phase or earlier data that shows that it's
4	reasonable to think of this collection of diseases as
5	potentially ones that would respond homogeneously to
6	the treatment effect. If you know something's
7	different and you're asking a separate question and
8	you shouldn't be including them in the model.
9	DAN RUBIN: All right, I think that's
10	going to close our moderated discussion. So, thank
11	you, everyone, for making this a great event. And our
12	next item on the schedule is Dr. John Rex is going to
13	provide a summary.
14	JOHN REX: So, let me first say that
15	this has been a really instructive day and a half.
16	And Sumathi and I put our heads together and thought
17	about the stuff we've learned the last day and a half
18	and where it might go. And I'm nominated to talk
19	about it, but this is really a group work product.
20	So, all of you are here as well.
21	So, these are the big messages we're
22	going to cover: AMR Enterprise in crisis. We can't

1 fix it all today but some things we can. What are the 2 emerging ideas? What do we do next? And so, the fact 3 that the AMR Enterprise is in crisis, that the Check 4 Engine light is flashing was nicely demonstrated by 5 several of the talks.

Late stage commercial failures have
occurred and seem likely to continue. And even when
successful, your stock gets shorted and your NPV goes
down, and the return is less than that of even a
moderately successful oncology product.

11 So, what are the elements that are 12 within our grasp? And this was really -- I learned 13 some things very helpful here. So, we know that push funding works, actually. CARB-X, BARDA, NIAD, 14 15 Wellcome Trust, Novo REPAIR. They have lit a bonfire in the preclinical space, and there are some neat 16 17 things coming. Can the science problem be solved? 18 Can we find new antibiotics? I think the answer is 19 It looks like we're going to. veah.

20 We've also learned that we can reliably 21 get products to approval -- many products to approval, 22 with basic studies in well-understood infections. And

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1	helpfully, these studies generalize reasonably well to
2	the U.S. We just had a good conversation about that.
3	Some more detail could be added here.
4	But, importantly, we can't generate the
5	same quality of data for all uses. And other possible
6	uses of a new drug are always going to exist and are
7	important to clinicians. Five o'clock this afternoon
8	you may have to treat a patient who's never going to
9	have a labeled drug for their meningitis. And they're
10	just pick any one of those things. There's nothing
11	labeled for it so what do you do?
12	So, this flips around to what do we
13	want? And broken down here, it's by important
14	stakeholder group. So, physicians and ID physicians
15	want access to all of the data, preferably interpreted
16	for them in some way. Payers and P&T committees would
17	like that same thing but they'd like a measure of the
18	quality of the data. Patients would like us to hear
19	their voice and think about how they feel net of the
20	whole process.
21	And companies would like validated and
22	acceptable mechanisms for promoting based on the data

1 on resistant pathogens and difficult infections. You
2 know, we get it -- the CUTI's not where you want to
3 wind up but it's hard to generate the rest of it. And
4 FDA had to label per regulation. And remember the
5 phrase, adequate and well-controlled.

So, two slides with ideas. The first 6 7 one is -- the broad title here is Tell the Story. 8 Theme A: Make clear to ourselves and our peers the 9 limits on data generation. Everybody needs to 10 understand what you can do, what you can't do. It's 11 very easy to wish for me, especially when you've never 12 -- you can't feel the complexity of producing the 13 information. And if there's a better way to generate this data we would be doing it. We've searched long 14 15 and far to find ways to generate the data and we are 16 constrained.

We as a community also have to be very clear on the limits on the product label. And without rules, we would have the Wild West, arbitrary decision making. And the role of the FDA is to consistent -one of their many roles is to consistently apply the rules so that we know what's coming. And the standard

of that is well-controlled, and that's the standard.
 Work with that.

3 But there is something else that I 4 hadn't anticipated was going to come out of this, and this was this idea of sharing the other available 5 Keeping in mind the limits on the label, we get 6 data. 7 the data published -- we should talk about the nuances of the trial. But there's something else that I think 8 we could be looking for, and that is a sort of peer-9 10 to-peer communication. We're nominating IDSA here to 11 the society for validating -- you can validate by 12 publishing an informed critique of the available 13 secondary data.

And I do mean this as peer-to-peer. 14 15 It's 4 o'clock, I'm stuck with this patient, I might send my fellow to the library. He or she might find 16 17 the right information. It's actually better if Helene 18 spent some time six months ago reviewing it and she 19 wrote down what she thought she would do in this unusual circumstance. It's the information that you 20 21 would like to have.

22

And reviews like that can be used for a

1	number of things. They can be used by a company for
2	discussion with payers. If written correctly, it's
3	also the thing that when you want to use something
4	off-label you can show to the insurer to say, shut up
5	and just pay for it. That's what's in this document.
6	Write it down in advance. And it might be usable for
7	promotion. I'm not sure about that or not. We also
8	need to include Europe in this conversation. So, this
9	is all about telling the story.
10	The other part is Use the Data. Theme
11	A: Be Clear on the Power of the Standard Indication.
12	Modern non-inferiority sites are powerful tools. They
13	do detect inferior agents, so it's an interesting
14	presentation from Aaron about how few it takes to find
15	the dog, you know? You're going to know pretty
16	quickly if it's no good. They provide clear safety
17	and efficacy comparisons, they facilitate initial
18	approval, and they provide the basis, the launch pad
19	to get you to the other stuff. And then better use of
20	the data we already have.
21	A bunch of neat ideas here that are
22	going to take some thinking. How do we borrow data

across indications? Could it be that we use different 1 2 thresholds for different settings in terms of the data that we're willing to validate? And don't forget 3 4 about the idea of (indiscernible) patient-oriented measures. And I'm going to add DOOR. It didn't get 5 typed in here in time. But that's the idea that we 6 7 should be working with. 8 And then, finally, generate the data more efficiently. I've heard some nice discussions 9 10 about that. It's not a panacea. Platform trials show 11 a potential to reduce cost and speed data generation.

12 The idea that the sites would already have a contract 13 in place. That right there saves a year in getting 14 the trial actually running.

And the trial platform thing seems particularly true after initial approval is achieved. Studies in pediatrics, HAP/VAP, and rare infections would be especially suitable here. That was a theme that we heard when developing the concepts that you've heard about for the master protocols.

21 So, it's the last slide. Great 22 conversation. And I want to give a shout out to

Sunita Shukla, who is sitting right there and was the
 ringmaster for making this meeting happen. Well done.
 Thank you very much for your help with that.

4

(Applause)

5 She had a bunch of people to chase down and she made it happen. Nothing is set in stone as to 6 7 what's going to happen next but a subsequent debate 8 seems needed on particularly the three topics listed How do we borrow data across indications? The 9 here: 10 idea of different thresholds for different settings --11 what is adequate and well? And how do we include 12 Europe in this conversation?

And I think if we start working -those are things we can work on, in addition to the things the societies can work on, and I'm hopeful that we will have a next conversation. So, thank you to FDA, thank you to IDSA, thank you to Pew, thank you to the NIH for convening this session. Safe travels. (Applause)

JOHN FARLEY: And I think I speak for all of us in the Department of Health and Human Services that have been part of this, that we have

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1	gotten a lot of very useful input and ideas. And I
2	also want to thank IDSA, and Pew, and NIH for co-
3	sponsoring this event. We're committed to support
4	some of the efforts that will come out of this meeting
5	and we're also committed to reconvening this group at
6	an appropriate time in the future because we think
7	this was a really useful meeting. So, thank you.
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