PRIVACY AND CIVIL LIBERTIES OVERSIGHT BOARD

Defining Privacy Forum

November 12, 2014

The public forum was held at the Georgetown

Marriott Hotel, 1221 22nd Ave, N.W., Washington,

D.C. commencing at 8:30 a.m.

Reported by: Lynne Livingston

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                       BOARD MEMBERS
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    David Medine, Chairman
    Rachel Brand
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    Patricia Wald
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    James Dempsey
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    Elizabeth Collins Cook
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                         SESSION 1
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                Defining Privacy Interests
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    Ed Felten, Princeton University
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    Liza Goitein, Brennan Center for Justice
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    Paul Rosenzweig, Red Branch Consulting, PLLC
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    Dan Solove, George Washington University Scool of
    Law
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3 1 SESSION 2 2 Privacy Interests in the Countertrrorism Context and the Impact of Technology 3 Annie Anton, Georgia Tech University 4 Alvaro Bedoya, Geogretown Center on Privacy & 5 Technology 6 Michael Hintze, Microsoft 7 Hadi Nahari, Chief Security Architect, NVIDIA 8 9 SESSION 3 10 Privacy Interests Identified and Addressed by Government Privacy Officials 11 Alex Joel, Office of the Director of National 12 Intelligence 13 Erika Brown Lee, U.S. Department of Justice 14 Rebecca Richards, National Security Agency 15 16 SESSION 4 17 Applying Lessons Learned from the Private Sector Fred Cate, Indiana University School of Law 18 19 Harley Geiger, Center of Democracy and Technology 20 Chris Inglis, Paladin Capital Group and former Deputy Director of NSA 21 22

4 1 PROCEEDINGS 2 MR. MEDINE: Good morning, and welcome 3 to the Privacy and Civil Liberties Oversight Board's public meeting on defining privacy. 5 It's 8:30 a.m. on November 12th, 2014, 6 and we're meeting in the west-end ballroom in the Washington Marriott Georgetown Hotel in 8 Washington, D.C. 9 This hearing was announced in the 10 Federal Register on October 21st, 2014. As 11 chairman, I will be the presiding officer. 12 All five Board members are present and 13 there is a quorum. The Board members are Rachel 14 Brand, Elisebeth Collins Cook, James Dempsey and 15 Patricia Wald. 16 I will now call the hearing to order. 17 All in favor of opening the hearing, please say 18 aye. 19 (Vote taken.) 20 MR. MEDINE: We will proceed. 21 So what is privacy? The right to be

left alone? A desire for independence of

- 1 personal activity? The right to make decisions
- ² regarding one's private matters? Space for
- intellectual development, anonymity or obscurity?
- 4 Freedom from public attention? Freedom from
- being observed or disturbed by others? Freedom
- from intrusion into one's solitude? Avoiding
- 7 public disclosure of private facts about
- yourself? Freedom from publicity which places
- 9 you in a false light? Freedom from appropriation
- of your name or likeness? Control of how one's
- personal information is collected and used?
- 12 Freedom from surveillance.
- These are just a few definitions that
- have been given to privacy in the past. I expect
- during the course of today's discussion that
- we'll hear others.
- The meeting today and the comments we
- 18 receive will inform the Board's approach to
- privacy issues within its statutory mandate.
- There will be four panels today. The
- 21 first will focus on defining privacy interests.
- The second will consider privacy interests in the

- 1 counterterrorism context and the impact of
- ² technology.
- Next we will hear from government
- 4 privacy officials regarding privacy interests
- ⁵ that have been identified and addressed. The
- 6 final panel will see how lessons learned from the
- ⁷ private sector can be applied in the
- 8 counterterrorism context.
- 9 Each panel will be moderated by a
- different Board member, and after the Board
- member poses questions other Board members will
- have the opportunity to pose questions.
- And afterwards, members of the audience
- 14 are invited to submit written questions. Peter
- Winn has cards and people can get a card from him
- and submit the questions, time permitting, for
- the moderator to pose to the panelists.
- I want to thank the panelists who have
- agreed to appear here today on this panel and
- others.
- I also want to note that we have a
- strict timekeeper, Joe Kelly, sitting in front,

- and so panelists are encouraged to keep their
- remarks brief so we can have a more extensive
- 3 discussion.
- We will be taking a lunch break between
- ⁵ noon and 1:15.
- Today's program is being recorded and a
- ⁷ transcript will be prepared and put on our
- 8 website at plcob.gov in a week or so.
- 9 Written comments from members of the
- public are also welcome and may be submitted
- through regulations.gov through the end of the
- 12 year.
- Finally, I want to thank the Board's
- staff, Sharon Bradford-Franklin, Shannon Wilson,
- 15 Simone Awang, Lynn Parker Dupree, Renee
- 16 Gewercman, Peter Winn, Joe Kelly for their
- efforts in making today's event possible.
- And so we'll now turn to the first panel
- 19 moderated by Judge Wald.
- MS. WALD: Thank you. Panel 1 will
- 21 attempt to explore, I think it would be too
- 22 ambitious to say define privacy, and the many

- separate individual and societal interests that
- the notion of privacy encompasses.
- The novelist Jonathan Franzen
- ⁴ perceptively remarked, "Privacy is like the
- ⁵ Cheshire cat of values, not much substance but a
- 6 very winning smile. Legally, the concept is a
- 7 mess." That's a quote.
- 8 That may be unduly pessimistic. Most
- 9 commentators do agree that there are aspects of
- 10 privacy that go way back to the most ancient
- 11 civilizations, and that our own Founding Fathers
- enshrined several of them in the Bill of Rights.
- But the concept of privacy has become a
- 14 receptacle for a conglomerate of interests or
- values that individuals and society care about,
- but which to varying degrees they're willing to
- balance with competing values, such as national
- 18 security.
- Thus, the law of privacy consists mainly
- of a series of situations in which courts,
- legislators or government officials have decided
- to recognize a privacy interest, or not to, and

- to protect, or not to, that interest against a
- ² competing value.
- 3 So our panelists today will identify the
- 4 varied individual and societal interests that
- 5 travel under the rubric of privacy and discuss
- 6 how far and under what conditions our laws do or
- ⁷ should legitimate claims that are based upon
- 8 those particular interests.
- Now our format will be for each panelist
- to talk initially for seven minutes. And the
- 11 gentleman in the front row will give you a yellow
- card two minutes before, and a green card will
- mean it's time to wrap up.
- And then at the end of their initial
- speeches, then I will question them as the
- moderator for about 20 minutes. Then that will
- 17 be followed by another 20 minutes of questions by
- my fellow Board members.
- 19 After that, and I hope there will be
- some time left for the written questions which
- 21 members of the audience are invited to send to
- the people who circulate to collect them, and

- then I will question them. I will discuss some
- of those questions with the people on the panel.
- You already, I think, have bios of our
- 4 illustrious panelists, but I'm going to identify
- 5 them very briefly before they speak.
- So we'll start off, Liza Goitein is a
- 7 Co-Director of the Brennan Center's Liberty and
- 8 National Security Program.
- 9 MS. GOITEIN: Thanks very much, Judge
- Wald. And I apologize in advance, I have a cold
- so my voice kind of comes and goes, but thank you
- to all of the Board members for inviting me to
- 13 participate in today's discussion.
- If there's one thing I've learned from
- my own involvement in privacy issues over the
- past few years is that privacy is different
- things to different people.
- David gave a very comprehensive list of
- some of the things that privacy is. I'm not sure
- what I would add to that, except to say that I
- think that for those who are outside the
- ideological mainstream in this country, privacy

- vis-a-vis the government can be critical to
- effectuate other rights, such as the freedom to
- ³ religion, speech and association.
- 4 So collectively as a society we value
- 5 all of those aspects of privacy, even if some of
- 6 us value only some of them, or none of them.
- So what does that mean for our analysis?
- 8 I think it's interesting for us to think about
- ⁹ different definitions of privacy, and it's
- helpful insofar as it shows the range of
- definitions that are out there.
- But I'm not at all convinced that
- 13 Congress, or the courts, or this Board should be
- in the business of attempting a granular
- definition of privacy or its importance.
- Look at the freedom of religion, by way
- of comparison. Courts don't probe what religion
- is or why it's important. And that's not because
- the definition of religion is obvious, by any
- means. It's at least in part because of the
- opposite, because religion is different things to
- different people.

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2 concept of religion that's broad enough to 3 encompass the many different roles that religion plays in people's lives, and then the court 5 protects it, except in the rare circumstance 6 where there's an overriding governmental interest. And Congress has followed the same 8 approach. 9 When it comes to information privacy, 10 which is what I focus on in my job, the best 11 working concept of privacy, the concept that best

So what the court does is it adopts a

This concept avoids to some extent the

privacy serves, is control of information.

encompasses all of the important interests that

- what and the why of privacy, and focuses instead
- on the how, how privacy is realized as a
- 17 practical matter.
- And it also has the additional advantage
- of matching up quite well with the text of the
- Fourth Amendment. If a person controls her
- papers, she is secure in them. If a person does
- not control her papers, she is not secure in

- 1 them.
- What are some of the ramifications of
- 3 this concept of privacy? Well first, controlling
- 4 one's information means controlling not only what
- one shares, but with whom and under what
- 6 circumstances.
- ⁷ I may share certain information with my
- 8 mother or with a close childhood friend, but that
- 9 doesn't mean that I've chosen to share that
- information with the entire world, including the
- ¹¹ NSA.
- Sure, there's a chance my mother might
- 13 rat me out. There's a chance that my childhood
- 14 friend has a tax problem I didn't know about and
- could be pressured by the government into
- becoming an informant.
- But to equate this outside risk that my
- 18 confidences may be misplaced, with a willing
- disclosure to everyone in the world is a legal
- fiction of the worst kind, and that's really what
- the third-party doctrine is, in my view.
- Second, you don't, in fact, relinquish

- all control over information about your public
- ² activities by virtue of walking out your front
- door. There is such a thing, functionally
- speaking, as privacy in public.
- 5 And this is something that's
- 6 well-understood in the FOIA context, the Freedom
- of Information Act context. There's a privacy
- 8 exception under FOIA which allows the government
- ⁹ to withhold information if releasing it would
- unduly compromise personal privacy. Think of
- 11 Social Security records.
- The Supreme Court held in 1989, that a
- 13 rap sheet would be covered by this exemption,
- 14 despite the fact that all of the information in a
- 15 rap sheet is available by virtue of a diligent
- door-to-door combing of court records.
- So why was the rap sheet still private?
- 18 Because the court held while the information in
- it was publicly available, it was practically
- obscure.
- This is such a commonsense concept and
- it deserves a home in Fourth Amendment

- jurisprudence. The sum total of a person's
- 2 movements in public over extended periods of time
- may be publicly available information, but using
- 4 normal powers of human observation it is
- ⁵ practically obscure.
- So when the government uses drones, or
- ⁷ stingrays, or GPS technology to pierce that
- 8 obscurity, it has compromised the control that
- ⁹ the person would otherwise exercise over this
- information, and that's a privacy violation.
- 11 Third, privacy violations happen at the
- point that the information is collected. We've
- 13 heard intelligence officials recently telling us
- that we don't have to worry about the NSA's
- 15 collection of bulk collection of telephone
- 16 records because nobody looks at the records
- unless they have reason to suspect some kind of
- 18 terrorist link. That is the government telling
- 19 you what aspects of privacy you should value.
- Many people won't care if the government
- 21 collects but doesn't look. Other people won't
- care if the government looks but doesn't

- ¹ prosecute.
- But the point at which the government
- 3 collects the information is the point at which
- 4 you've lost control. And for plenty of people
- 5 that loss of control itself produces harm. It
- 6 produces a feeling of vulnerability. It causes
- ⁷ people to change their behavior.
- In 2014, there was a poll after the
- 9 Snowden disclosures showing that 47 percent of
- respondents had changed their online behavior
- 11 after those disclosures.
- There was another survey of 520 American
- writers showing that one out of six authors,
- 14 after the Snowden disclosures, refrained from
- writing about certain topics because they feared
- ¹⁶ surveillance.
- After news stories broke about the
- 18 NYPD's infiltrations of Muslim student
- 19 associations, attendance in those associations
- ²⁰ dropped.
- In some ways these are some of the worst
- harms that come from privacy violations because

- they're society-wide. They impact the way we
- function as a society. They impoverish our
- 3 social discourse by causing people to sensor
- 4 themselves and not put ideas out there.
- One last ramification of this concept of
- ⁶ privacy -- if I have time, I can't believe I have
- ⁷ time -- is young people. So I hear it said quite
- 8 often that young people don't care about privacy.
- 9 And it's certainly true that many young people go
- on Facebook and share incredibly personal
- information with 622 friends. But they don't
- share that information with 623 friends.
- What they share and the number of people
- that they share it with may very well have
- changed, it certainly appears so, but they still
- 16 control the sharing, or at least they think they
- 17 do.
- And my impression, based on a totally
- unscientific survey of all the young people in my
- life, is that they still value that control.
- So, the red card, I knew it was coming.
- 22 All right, I'll stop there.

- MS. WALD: Okay, thank you. Professor
- ² Daniel Solove is the John Marshall Harlan
- Research Professor of Law at the George
- Washington Law School.
- MR. SOLOVE: Good morning. I would like
- to make five brief points this morning.
- 7 The first point is that privacy is much
- 8 more than hiding bad secrets. One of the common
- ⁹ arguments that people often make about privacy is
- that people shouldn't worry if they have nothing
- to hide. And I hear this argument all the time.
- This argument, and many other arguments
- about privacy, are based on a conception of
- privacy, a conception of privacy that's very
- narrow, that sees privacy as hiding bad or
- discreditable things.
- Well, privacy is much more than that.
- Privacy isn't just one thing, it's many different
- things. Privacy involves keeping people's data
- secure. It involves the responsible use of data.
- It involves making sure that when data
- is kept, it's kept accurately. It's making sure

- that people who keep the data are responsible
- stewards of that data, that people have rights in
- that data and some participation in the way that
- ⁴ data is used.
- 5 All these things have nothing to do with
- 6 nothing to hide. They have nothing to do with
- 5 secrets and everything to do with how their
- information is kept, collected, stored, etcetera.
- I think that if we see privacy broadly
- we can move away and abandon these very narrow,
- 11 cramped views of privacy.
- The second point I'd like to make is
- that privacy is a societal interest, not just an
- 14 individual one.
- When balancing privacy and security,
- 16 privacy is often seen as an individual right and
- then security is often seen as a social right.
- 18 And when they're balanced, society generally wins
- out over the individual. And I think this
- 20 actually skews the balance to the society side,
- 21 to the security side.
- But, in fact, privacy isn't just an

- individual interest. It doesn't just affect the
- individual, it's a societal interest. We protect
- privacy because we want to protect society. We
- 4 want to shape the kind of society we want to live
- ⁵ in.
- Privacy doesn't just protect the
- ⁷ individual for the individual's sake, it protects
- 8 the individual for the society's sake, because we
- ⁹ want a free society where people are free to
- think and speak without worrying about negative
- 11 consequences from that.
- The third point I'd like to make is that
- the collection of personal data through
- 14 surveillance and other means of government
- information gathering can cause significant
- problems.
- Data collection and surveillance aren't
- inherently bad, but just as industrial activity
- 19 causes pollution, government surveillance and
- data gathering can cause problems. And these
- 21 problems must be mitigated. They must be
- 22 addressed when they clash with important

- ¹ interests.
- Some of the problems include, one, that
- this activity can chill people's expression. It
- 4 can chill people's exploration of ideas. It can
- 5 chill people in many different ways. Either they
- 6 might not say something, or they might say
- ⁷ something slightly differently, or they might act
- 8 differently, or do things differently. We don't
- ⁹ want that chilling when it comes to legal
- 10 activity.
- The other thing, the other problem, is
- that surveillance gives a lot of power to the
- watchers. There's a lot of things that can be
- done with a vast repository of data beyond a
- particular aim that it might have been collected
- for. Data has a way of often being used in other
- manners, in other ways.
- I think that another issue too is the
- level of accountability and oversight that goes
- into this, because it's about the structure of
- our government and the relation of the government
- to the people that we're talking about here.

- What kind of accountability will the
- ² government have when it gathers all this
- information? What limits will there be on the
- information gathered and used? How long will the
- ⁵ information be kept?
- In a free society people are free to act
- as they want to act, as long as it's within the
- bounds of the law without having to justify
- ⁹ themselves.
- They don't have to go and explain their
- actions to a bureaucrat sitting in a room full of
- television monitors about what they're doing.
- 13 They don't have to go and explain themselves when
- 14 a computer's lights are blinking red because of
- something that they said and it can be
- misinterpreted.
- People don't have to worry about that.
- 18 They can act freely without having to worry about
- 19 how suspicious their actions might look. That is
- a key component to freedom.
- The fourth point I'd like to make is
- that we can't adequately balance privacy and

- security without a reasonable amount of
- ² transparency.
- There's an overarching principle that
- 4 this nation was founded upon, it is that we the
- people are the boss. The government is our
- 6 agent. We can't evaluate what government
- officials are doing if we don't know what's going
- 8 on.
- Now this doesn't mean there should be
- absolute transparency, but it does mean that we
- need to know something, enough to be able to
- evaluate government surveillance.
- Because ultimately the choice about the
- 14 proper level of surveillance isn't the NSA's to
- make, it's not the President's to make, it's the
- people's choice. We can't forget that. It's the
- people's choice, and the people must be given
- sufficient information to make that choice.
- My last point is that the government
- 20 must get buy-in from the people for its
- 21 surveillance measures. Without buy-in, people
- are going to start to take self-help measures,

- which is something that we see happening now.
- We see that companies are providing
- people with ways to encrypt their data to protect
- 4 it from snooping government entities. This is
- ⁵ the market speaking. This is something that
- 6 people want. This is something being sold to
- people that people are going to buy. This is
- 8 something in demand.
- Why? Why are people demanding this?
- 10 Because they've lost trust, because the laws
- 11 regulating government surveillance are weak and
- do not provide adequate oversight or
- 13 accountability.
- 14 This is why strong privacy protections
- aren't necessarily bad for security. In fact,
- they ensure that the people are comfortable that
- there is adequate oversight and accountability
- 18 for that surveillance and that they're
- comfortable and know that they have the
- information that they need to continually
- evaluate what's going on.
- And if they can evaluate what's going on

- and buy into what's going on, things will be a
- lot better when it comes to balancing privacy and
- 3 security. Thank you.
- MS. WALD: Paul Rosenzweig is the
- ⁵ founder of the Red Branch Consulting Program, and
- a senior advisor to the Chertoff Group, and he
- was formerly Deputy Assistant Secretary for
- ⁸ Policy at the Department of Homeland Security.
- 9 MR. ROSENZWEIG: Thank you, Judge Wald,
- and thank you, Mr. Chairman and members of the
- Board. I appreciate the opportunity to speak
- with you today.
- 13 It's really entirely appropriate for the
- Board to begin a discussion of privacy in this
- new technological age. In fact, in my judgement
- it's essential.
- And the reason for that is essentially
- one that puts me in some disagreement with my
- 19 fellow panelists. I think that our conceptions
- of privacy, founded as they were back in the
- ²¹ 1970s with the FIPPs, are somewhat outdated and
- 22 antiques that don't survive the technological

- 1 challenges that we face.
- The 1973 Thunderbird was a marvelous car
- but we don't think of holding it out today as the
- 4 state of automotive engineering. Nor would I
- 5 think we should address the FIPPs as the state of
- ⁶ privacy thinking. We need, in effect, a Tesla
- ⁷ for privacy today.
- What would that look like? Well, there
- 9 are many ways to answer that question, and I
- think to answer it you have to begin by thinking
- about what sort of value privacy is.
- And here, again, I think I find myself
- in some disagreement with other members on the
- panel and perhaps with members of the Board. I
- do not think that privacy is an ontological
- value. I don't think it's akin to religion.
- 17 It's not an inherent human right or the product
- of some natural law.
- Rather in my judgment privacy is an
- inherently instrumental value, one that acts in
- the service of other societal values. It's a
- utilitarian value that derives its worth only

- insofar, in my judgment, as it fosters other
- ² positive social gains.
- Privacy for its own sake is just an
- assertion of autonomy from society. It is
- ⁵ valuable insofar as it advances other objectives.
- Now let me kind of put some salt on
- ⁷ that. The problem is that buried in the word
- 8 privacy are many different social values that
- 9 we're fostering, too many really to catalogue,
- though the chairman did a good job of trying to
- 11 start.
- For example, we often see in the
- discussion here privacy is enhancing our freedom
- 14 from government observation. That's probably the
- use that's most salient to what the Board does.
- But it also enables democracy. That's
- why we keep the ballot private. It fosters
- 18 personal morality. That's why we keep the
- 19 confessional private.
- Privacy is also about restraining
- government misbehavior, which is why we see
- 22 privacy values in the Fourth Amendment and other

- 1 procedural limitations on government action,
- another way in which privacy is obviously
- ³ relevant to this Board.
- 4 And it's also, as Dan said, sometimes
- 5 about transparency in the sense that we have
- 6 privacy rules so that I know what you know about
- 7 me.
- It can be about control, about control
- 9 of my own image.
- And it's sometimes also about simply
- shame, since one ground of privacy is enabling me
- to keep from the world things that I'm not proud
- 13 I did, of which there are far too many, I fear.
- What's important to note is that in all
- of these instances the value that we're
- protecting that underlies privacy is different
- 17 from the privacy itself.
- And that in turn suggests to me that the
- way to think about privacy is to think about what
- operational activities would protect the
- 21 underlying value most.
- It means we need to go to a micro level

- to understand in general the nuance that arises
- from the particular interest that is at the core
- of the privacy that we're talking about.
- For example, we protect the
- 5 confidentiality of attorney client
- 6 communications. Why? Because we think we need
- ⁷ to foster candor in the discussion between a
- 8 client and an attorney. That's something that we
- ⁹ feel so strongly about that the instances in
- which we permit that privacy to be violated are
- 11 few and far between, and they come only with the
- 12 highest level of judicial scrutiny.
- The Fourth Amendment itself reflects a
- similar utilitarian value of the security of our
- persons, places and things against intrusion.
- Once again, we impose a high bar, a probable
- cause requirement and a strong independent
- outside adjudicator, a judge issuing a warrant.
- But those aren't the only mechanisms by
- which we can protect privacy. We have a series
- of administrative processes that are often
- adequate to protect and restrain government

- ¹ observation.
- They're embedded in many of the internal
- 3 reviews that are very common in the IC, in the
- 4 intelligence community that you spend your time
- ⁵ reviewing.
- They're common in virtually every
- ⁷ institution of government that we have, at least
- 8 at the federal level that I'm familiar with,
- ⁹ where we think that administrative review,
- internal oversights, inspectors general,
- intelligence committee oversight are adequate
- 12 alternate administrative mechanisms.
- So what does that mean for some of the
- things that you think about? Let me look at the
- two programs that you've written about and just
- kind of express something there.
- The 215 program is one that directly
- impacts issues of government abuse or potential
- abuse because of the pervasiveness of the
- collection that underwent, that was there.
- It strikes me that that sort of
- pervasive collection is one that would require a

- strong, independent review mechanicism because of
- the comprehensiveness of its activity.
- By contrast, the 702 program, which
- 4 seems from what I've read from the outside from
- your reports, more narrowly focused, is one in
- 6 which less error correction mechanisms are
- necessary, less likelihood of inadvertent abuse
- 8 is there.
- 9 So if you press on what is being
- protected, you get a sense of a better way to
- 11 protect it.
- Let me say one brief word more about
- 13 transparency. I completely agree with others on
- the panel that transparency is essential to
- 15 control conduct and misconduct.
- But the critical question is, what type
- of transparency? And for me, again, this
- 18 requires us to ask what transparency is for.
- 19 It's the ground of oversight and audit.
- ²⁰ Transparency without that ground is just
- voyeurism.
- But absolute transparency, as Dan said,

- can't be squared with the need for secrecy in
- ² operational programs.
- I sometimes think that some calls for
- 4 transparency, though I hasten to say not by any
- other members of the panel or on the Board, are
- 6 really just coded efforts to discontinue
- ⁷ surveillance programs altogether.
- 8 The truth is that if we believe in
- 9 absolute transparency, we've gone a long way to
- the view that democracies can't have secrets, a
- view which I think is untenable in the modern
- world.
- And with my last thirty seconds let me
- offer one last thought about the role of the
- 15 Board and the multi-varied nature of privacy.
- Because I think that privacy is many
- things and has many applications in many
- different contexts, I also think that the most
- appropriate ground for making judgements about
- privacy is not in boards or judiciaries, but in
- the most representative bodies that we have
- available to us, in this instance Congress.

- I realize that's perhaps leaning rather
- 2 heavily on a body that is not held in the highest
- ³ regard at this time, but nonetheless, that is the
- 4 mechanism in a democracy for accumulating diverse
- ⁵ preferences, weighing them in the balance and
- for a broader societal
- ⁷ interest.
- MS. WALD: Okay.
- 9 MR. ROSENZWEIG: Thank you. My
- ¹⁰ apologies.
- MS. WALD: Ed Felten is a Professor of
- 12 Computer Science and Public Affairs at Princeton
- and founder of the Princeton Center for
- 14 Information Technology Policy. So he'll give us
- a somewhat different lens through which to view
- 16 privacy.
- MR. FELTEN: Thanks for the opportunity
- 18 to testify. Today I'd like to offer a
- 19 perspective as a computer scientist on changing
- data practices and how they've affected how we
- think about privacy.
- We can think of today's data practices

- in terms of a three stage pipeline. First,
- ² collect data, second, merge data items, and
- third, analyze the data to infer facts about
- 4 people.
- 5 The first stage is collection. In our
- 6 daily lives we disclose information directly to
- people and organizations. But even when we're
- 8 not disclosing information explicitly, more and
- 9 more of what we do online and off is recorded.
- And online services often attach unique
- identifiers to these recordings which are used to
- 12 link them up again later.
- The second stage of the pipeline merges
- the data. If two data files can be determined to
- correspond to the same person, for example,
- because they both contain the same unique
- identifier, then those files can be merged.
- And merging can create an avalanche
- 19 effect because merged files convey more precise
- information about identity and behavior, and that
- 21 precision in turn allows further merging.
- One file might contain detailed

- information about behavior and another might
- ² contain precise identity information. Merging
- those files links behavior and identity together.
- The third stage of the pipeline uses big
- 5 data methods such as predictive analytics to
- 6 infer facts about people.
- One famous example is when the retailer
- 8 Target used purchases of a product such as skin
- 9 lotion to infer pregnancy.
- Today's machine learning methods often
- enables sensitive information to be inferred from
- 12 seemingly less sensitive data.
- 13 Inferences also can have an avalanche
- effect because each inference becomes another
- data point to be used in making further
- ¹⁶ inferences.
- 17 Predictive analytics are most effective
- in inferring status when many positive and
- 19 negative examples are available. For example,
- Target used many examples of both pregnant and
- 21 non-pregnant women to build its predictive model.
- By contrast, a predictive model that

- tried to identify terrorists from everyday
- behavioral data would expect much less success
- because there are very few examples of known
- 4 terrorists in the U.S. population.
- With that technical background let me
- discuss a few implications for privacy. First,
- ⁷ the consequences of collecting a data item can be
- 8 very difficult to predict. Even if an item on
- 9 its face doesn't seem to convey identifying
- information, and even if the contents seem
- harmless in isolation, the collection could have
- 12 substantial downstream effects.
- We have to account for the mosaic
- effect, in which isolated, seemingly unremarkable
- data items combine to paint a vivid and specific
- picture. Indeed, one of the main lessons of
- 17 recent technical scholarship on privacy is the
- 18 power of the mosaic effect.
- To understand what follows from
- 20 collecting an item we have to think about how
- that item can be merged with other available
- data, and how the merged data can in turn be used

- to infer information about people. We have to
- take into account the avalanche effects that can
- occur both in merging and inference.
- For example, the information that the
- ⁵ holder of a certain loyalty card account number
- 6 purchased skin lotion on a certain date might
- ⁷ turn out to be the key fact that unlocks an
- 8 inference that a particular identifiable woman is
- ⁹ pregnant.
- Similarly, phone call metadata, when
- 11 collected and analyzed in large volume has been
- shown to enable predictions about social status,
- affiliation, employment, health and personality.
- The second implication is that data
- handling systems have gotten much more
- complicated, especially in the merging and
- analysis phases, that is the phases after
- 18 collection.
- The sheer complexity of these systems
- 20 makes it very difficult to understand, to predict
- 21 and to control how they behave. Even the people
- who build and run these systems often fail to

- understand fully how they work in practice, and
- this lead to unpleasant surprises, such as
- 3 compliance failures or data breaches.
- 4 Complexity frustrates oversight, it
- ⁵ frustrates compliance and it makes failure more
- 6 likely. Despite all best intentions
- organizations will often find themselves out of
- 8 compliance with their own policies and their own
- 9 obligations. Complex systems will often fail to
- 10 perform as desired.
- 11 Complex rules also make compliance more
- difficult. It's sometimes argued that we should
- abandon controls on collection and focus only on
- regulating use. Limits on use do offer more
- 15 flexibility and precision in theory, and
- sometimes in practice.
- But collection limits have important
- advantages, too. For example, it's easier to
- comply with a rule that limits collection than
- one that allows collection and then puts
- 21 elaborate limits on usage afterward. And
- 22 collection limits make oversight and enforcement

- ¹ easier.
- Limiting collection can also nudge
- 3 agencies to develop innovative approaches that
- 4 meet their analytic needs, while collecting less
- ⁵ information.
- The third implication is the synergy
- between commercial and government data practices.
- As an example, commercial entities put
- ⁹ unique identifiers into most website accesses.
- 10 An eavesdropper collecting traffic can use these
- identifiers to link a user's activity across
- different times and different online sites, and
- an eavesdropper can connect those activities to
- identifying information.
- Our research shows that even if the user
- switches locations and devices, as many users do,
- ¹⁷ an eavesdropper exploiting commercially placed
- identifiers can reconstruct 60 to 75 percent of
- what a user does online and can usually link that
- data to a user's identity.
- My final point is that technology offers
- 22 many options beyond the most obvious

- technological approach of collecting all of the
- data, aggregating it in a single large data
- 3 center and then analyzing it later.
- 4 And here I think Paul's analogy to the
- ⁵ 1973 Thunderbird is a good one. We would no
- 6 longer accept the safety technologies that were
- available on that vehicle. Nowadays we expect
- 8 airbags, we expect anti-lock brakes, we expect
- ⁹ crumple zones. We expect the latest technology
- to be used to make the technology safer and to
- 11 reduce risk.
- 12 And we should ask for the same when it
- comes to privacy. We should ask agencies to use
- 14 advanced technologies to limit how much
- information they collect, to use cryptography to
- limit undesirable flows of information.
- There's a large and growing literature
- on privacy-preserving data analysis and methods.
- 19 Determining whether collection of particular data
- is truly necessary, whether data retention is
- truly needed and what can be inferred from a
- 22 particular analysis, these are deeply technical

- ¹ questions.
- In the same way that the Board asks
- ³ probing legal and policy questions of the
- 4 agencies you oversee, I hope you'll build a
- 5 capacity to ask equally probing technical
- ⁶ questions.
- 7 Legal and policy oversight are most
- 8 effective when they're combined with
- 9 sophisticated and accurate technical analysis,
- and many independent technical experts and groups
- 11 are able and willing to help you build this
- 12 capacity.
- 13 Thank you for your time and I look
- 14 forward to your questions.
- MS. WALD: Thank you. Okay, for the
- next 20 minutes or so I'm going to pose some
- questions to the members of the panel, and I'll
- 18 pose them to a particular member, but then if one
- of the other members has something very cogent,
- 20 as I'm sure everything you say is cogent, feel
- 21 free to contribute.
- Liza, I'm going to start with you. Our

- 1 Constitution defines certain aspects of privacy
- in the Fourth Amendment, security of one's home
- 3 and papers from unreasonable search and seizure
- ⁴ and protection from general warrants.
- 5 But are there other aspects of privacy
- that the advocacy community believes deserve
- ⁷ legal recognition and judicial oversight, or can
- 8 they all be encompassed within the bounds of the
- ⁹ constitutional quarantees?
- And if so, what are the ones you think
- ought to be specifically recognized, protected in
- 12 our law?
- MS. GOITEIN: Sure. Okay, so to start
- with I suppose the obvious, the Fourth Amendment
- applies only to the government. It's a
- 16 restriction on the government. It's not a
- 17 restriction on private parties.
- And I think there's absolutely a place
- 19 for regulation of private entities and how they
- 20 control, acquire and control people's
- information. Because the market doesn't always
- do a great job of many things, although it does a

- ¹ great job of other things.
- But we certainly know that people are
- not a hundred percent satisfied with the privacy
- 4 protections that have been provided in the
- ⁵ private sector, and that obviously falls outside
- of the Fourth Amendment, but is deserving of
- ⁷ regulation.
- MS. WALD: While I've got you there
- ⁹ before you go on, there was another question that
- 10 follows directly from this.
- We hear an awful lot about the
- commercial acquisition of so much personal
- information and what they do with it. And in
- 14 fact, the argument is sometimes made, look, don't
- worry so much about the government, but some of
- the private, Google, some of the communications,
- the Internet, have great masses of data.
- Do you think that there's any
- significant difference in the risks to privacy
- that are displayed by the holdings of so much
- personal information by the government, as
- opposed to private entities, or is it like two

- big behemoths?
- MS. GOITEIN: I do think there's a
- difference. I think that difference may be
- 4 getting smaller, but I think there is a
- ⁵ difference. There remains a difference, which is
- that private companies do not have the same
- 7 coercive power over the individual that the
- government has, and private companies and private
- 9 entities don't have the same motivations to
- persecute people based on ideology or religion.
- I mean these are things that we have
- seen in the history of this country,
- 13 unfortunately. We have seen people targeted for
- surveillance because they were political enemies
- of the reigning administration.
- So what I would say is that private
- entities have neither the ability nor the motive
- to throw people in jail on pretext because they
- ¹⁹ are politically opposed to the current
- ²⁰ administration.
- That said, I think companies, the line
- between big companies in this country and

- 1 governments is getting thinner and thinner. And,
- you know, certainly companies might have some
- political axes to grind with respect to the
- 4 workforce, and they certainly have access to
- ⁵ people's information.
- I am not in the least bit unconcerned
- yith the private accumulation of information, but
- 8 I remain more concerned with privacy vis-a-vis
- ⁹ the government.
- MS. WALD: Okay. Let me try Professor
- 11 Solove.
- Now you wrote something in an article
- called, Conceptualizing Privacy, and you went
- into it a little bit in your prior remarks, there
- are sixteen kinds of activities that represent
- 16 privacy risks. Privacy itself has six aspects.
- 17 They're all defined too broadly and they're all
- defined too narrowly.
- And so you concluded, I think, if I read
- it correctly, that we should concentrate on
- 21 specific types of disruptions to those interests
- 22 and what should be done about that.

- Can you apply that kind of framework to
- the kinds of protection that we need in national
- 3 security data and surveillance programs, in
- 4 collection processing, identification, secondary
- ⁵ use, all of the other things that you talked
- 6 about in your article?
- 7 MR. SOLOVE: Yes, actually in what I
- 8 wrote I talked about privacy not being just one
- ⁹ thing and having a common denominator, but being
- a pool of common characteristics and actually
- applying to what I laid out was a taxonomy of
- 12 privacy about various types of problems.
- And I wanted to focus on the problems or
- 14 areas where certain activities cause disruption,
- they cause problems. And we want to mitigate
- those problems.
- And what are those problems? Because
- that's where we want to step in and say, hey, we
- should regulate this, we should do something
- about this, we should address these problems.
- It doesn't mean that the activities that
- cause them are bad, but it does mean that they do

- $^{
 m 1}$ cause the problems we need to address.
- Some of these problems that relate to
- ³ government data gathering include, one is
- 4 aggregation, that you can take a lot of different
- ⁵ pieces of data, each one being particularly
- innocuous, not really saying a whole lot about
- somebody, but when you combine them together, you
- 8 can learn new facts about somebody. This is what
- 9 data mining is all about, and data analytics.
- The whole becomes greater than the
- 11 parts. It starts to create a mosaic, a portrait
- of somebody.
- This then leads to the revelation of
- information that someone might not have expected
- or wanted when they gave out little pieces of
- information here and there.
- And I think this causes a problem. It
- disrupts people's privacy expectations. It can
- lead to knowledge of information that people
- don't want exposed or that society might not want
- 21 exposed. And so I think we need to address that
- 22 problem.

- And oftentimes conceptions of privacy
- will ignore aggregation because they'll say,
- well, if the information all were different facts
- 4 that were gathered from public information
- 5 there's no privacy.
- But I don't think that's true. I think
- ye really want to look at what the problems are
- 8 and if we look at the problems, there's a problem
- 9 here.
- Another aspect is a problem I call
- exclusion, which is the fact people lack an
- ability in a lot of cases to have any say in how
- that information might be used against them, any
- 14 right to correct that information, or to make
- 15 sure that it's accurate.
- And I think that's a key component of a
- 17 lot of privacy laws is there's a right for people
- 18 to make sure that proper decisions are being made
- about them based on their information.
- I can't go through all sixteen. I can
- hit some others.
- One is identification, the fact that

- this involves linking a body of data, what I call
- ² a digital dossier to a particular individual. By
- ³ identifying them you actually are connecting them
- 4 to data that then can be used to make decisions
- 5 about their lives. Some of the decisions could
- be good, but some decisions could in fact be
- 7 harmful to an individual.
- 8 Security is another issue that I see as
- 9 related and part of my taxonomy of privacy, and
- that's keeping data secure. When data isn't kept
- secure, it creates risks and vulnerabilities to
- 12 people that could expose them to a lot of harm
- if, in fact, the data is leaked improperly.
- And that happens all the time. We're
- all at risk when all this data is gathered
- together in a big repository.
- There are a lot of other things, but
- 18 I'll stop here in the interests of time.
- But these are just some of the ways that
- the taxonomy addresses this problem. I think
- it's important to think of the overarching point
- is don't start with some platonic concept of

- privacy and see, you know, what fits in it and
- what doesn't.
- I think it's better to look at things
- from the bottom up and say, where are the
- 5 problems here? What are the problems and harms
- 6 that are caused by these activities and how do we
- ⁷ address those harms?
- 8 MR. ROSENZWEIG: Could I just make a
- 9 brief comment?
- MS. WALD: Yes, sure.
- MR. ROSENZWEIG: I would agree with
- everything that Dan said but I would also say
- also look at what the benefits are.
- You know, the President's report on big
- data looked at the increase in the volume,
- velocity and variety of data, and championed the
- idea that large scale data aggregation creates
- ubiquitous new knowledge -- serendipitous new
- 19 knowledge that is of value to society as well.
- So it brings with it harm, but it also
- 21 brings with it benefits, and that is why I see it
- 22 as a kind of cost benefit utilitarian analysis.

- MS. GOITEIN: Sorry.
- MS. WALD: Yes, go ahead.
- MS. GOITEIN: I just want to say one
- thing quickly. I thought this it's a utilitarian
- ⁵ value, not a human right, it is a human right.
- 6 It's listed in the Universal Declaration of Human
- ⁷ Rights, it's listed in the ICCPR, and other
- 8 treaties and protocols that the United States has
- 9 signed and that have the force of customary
- 10 international law.
- So whatever one's personal feelings
- about that, I don't think this Board has the
- 13 latitude to decide that all these treaties we've
- signed declaring it as a human right are void.
- MS. WALD: Of course defining what's
- included in that human right has been one of our
- problems, it's been one of your problems, it's
- been everybody else's problem.
- MS. GOITEIN: Of course, but it's a
- ²⁰ human right.
- MR. SOLOVE: May I make one small point?
- MS. WALD: Yes.

- MR. SOLOVE: And that is I think I
- totally agree about the benefits of big data and
- the use of these things, but I think often the
- balance is wrongly cast between, okay, here,
- let's take the benefits and let's weigh it against
- 6 the harms.
- Because protecting privacy doesn't mean
- getting rid of big data, or not engaging in
- ⁹ surveillance, or not doing a search. The Fourth
- 10 Amendment allows searches and allows
- surveillance, for example, it just requires
- 12 certain oversight.
- So what we need to look at when we're
- balancing is not all the benefits of big data
- against privacy, we need to look at to what
- extent do oversight, accountability and these
- protections on it, to what extent do they
- diminish some of those benefits.
- And that difference, that diminishment
- is what gets put on the scale against privacy,
- 21 not all of big data's benefits. And I think if
- we weigh that appropriately, then I think we get

- ¹ a better balance.
- MS. WALD: All right.
- MS. GOITEIN: Very quickly.
- 4 MS. WALD: Briefly.
- MS. GOITEIN: Yes. We don't get to
- 6 weigh these things de novo when it comes to the
- ⁷ Fourth Amendment. The balance has been struck.
- ⁸ The government can't say we want to do searches
- ⁹ in people's houses, we have a really good reason,
- we don't have a warrant, but we have a really
- good reason, let's everybody do this balance
- 12 anew.
- 13 That balance was struck by the drafters
- of the Fourth Amendment. You need, in the vast
- majority of cases, there are some narrow,
- delineated exceptions, but you need a warrant
- based on probable cause of criminal activity to
- do those searches. This is not starting from
- 19 scratch.
- MS. WALD: Mr. Rosenzweig, your
- 21 approach, and you talked a little bit about this,
- your approach for balancing privacy and national

- security has I think been termed, whether you
- ² call it instrumental or consequential, but in one
- of your articles you talked about you thought
- 4 limiting the right of somebody to complain or to
- ⁵ go to court, etcetera, to intervening on the
- basis of when they are suffering a tangible harm,
- ⁷ like a warrant or being called before the grand
- ⁸ jury, as opposed to Professor Solove's views of
- ⁹ privacy as a kind of foundational value,
- 10 recognizable in its own right.
- 11 Yet you also recognize in some of your
- other works the significance of some aspects of
- 13 privacy to a democratic society.
- Now all of you have talked about it
- isn't just an individual right, it's a right that
- an open society needs, starting with even the
- 17 necessity for people developing their
- 18 personalities in an atmosphere in which they feel
- 19 free to experiment a little bit, to have
- 20 relationships, to talk without feeling that
- they're constantly being judged by the government
- or society.

1 I'm wondering how you reconcile your 2 recognition of the aspects of privacy that are 3 necessary to a democratic open society with this notion that we really shouldn't start intervening 5 until somebody is suffering some tangible harm. 6 MR. ROSENZWEIG: Well, thank you for the 7 I think that I don't see them as question. 8 irreconcilable because I see the question about 9 the adversity of consequence and the error 10 correction mechanisms as critical to the first 11 part of your question, the inherency of the 12 value. 13 To say that, I sort of sometimes use a 14 thought experiment, which is, what if in some 15 hypothetical world, which I assure you does not 16 exist, the government never abused anybody, never 17 actually misused the data it was collecting, 18 never, had no lists of enemies, no persecution, 19 and never made a mistake. 20 Now granted, that's an impossible 21 standard, but if that were the instance then in 22 the long run the values that underlie, the

- democratic values, would be supported and people
- would no longer fear the collection because the
- lack of adverse consequence, or by hypothesis, a
- 4 hundred percent would have gone away.
- So to my mind, the way to support the
- of values that we see in the underlying democratic
- ⁷ sphere is to build the error correction
- 8 mechanisms, the audits, the oversights, this
- 9 Board, into the process in a way that reassures
- society, as much as possible, that we're driving
- down the errors, and frankly both types, the
- 12 false positives and the false negatives, but this
- 13 Board is principally concerned with the false
- 14 positives, driving down the errors as much as we
- 15 humanly can.
- We don't eliminate government programs
- because of the possibility of error because every
- 18 government program, every human endeavor has the
- 19 possibility of error.
- We arm police officers, even though we
- 21 know that they will sometimes misuse their
- weapons. We don't eliminate that. We try and

- drive down the error rate as much as possible so
- that we engender people's confidence in the
- ³ police.
- 4 And we see sadly these days exactly what
- 5 happens when people's confidence in the police is
- 6 not maintained, that our error correction
- mechanisms are deemed by society inadequate.
- And I think we're sort of seeing some of
- ⁹ the same thing in response to the Snowden
- disclosures as well. But that suggests to me
- that the right way to support the underlying
- values is to go back and think about how to fix
- the error correction mechanisms.
- MS. WALD: So let me just pursue one
- thing that you brought up earlier, which has come
- up in some of our past reports and is bound to
- come up in future ones, I think, and that is at
- what stages, if you would go into a little bit
- more into the point at which you think an
- independent review of decisions outside of the
- 21 government's internal auditing and processes are
- necessary to ensure that you have this kind of

- trust by the people that the government is not
- taking risks to its privacy, and in terms of what
- you yourself suggested that history has got some
- lessons for us on the "trust us" aspect.
- MR. ROSENZWEIG: Well, I certainly don't
- dispute that we've had failures in the past.
- ⁷ Anybody who would dispute that hasn't read
- 8 history.
- I would say that there's no one size fit
- all answer, that it really depends upon the harms
- involved and the nature of what you anticipate
- the failure mode would be.
- 1'll give you two examples. On the one
- side we have the current TSA inspections programs
- at the airport. Probably a fairly significant
- error rate of false positives, pulling people
- ¹⁷ aside for secondary inspection.
- On the other hand, a comparatively
- 19 modest intrusion. And I say that knowing that
- 20 many people think it's a very large intrusion,
- but nonetheless comparatively modest compared to
- the coercive nature of being put in jail, for

- 1 example.
- So in that instance we seem reasonably
- 3 happy with a principally administrative
- 4 methodology that doesn't require any outside
- 5 check because individual liberty is not at issue,
- 6 long-term confinement is not at issue. The
- 7 degree of harm is small.
- By contrast you will infer that I
- 9 certainly think that independent review is
- essential whenever people's liberty is at stake,
- when significant aspects of livelihood are at
- 12 stake.
- I think that one of the strangest things
- that I see in the privacy debates today is that
- we seem to get all wrapped up about things like
- the TSA screening and we don't look at how
- government databases are used to deny employment
- to people.
- You can't get a job in the
- transportation industry with a record, even if
- the record is itself ripe with error, because of
- the transportation worker identity card.

1 So we, I think, have it backwards a 2 And that would be an instance where an 3 independent review of some sort for somebody who's denied employment in the nuclear industry 5 or in the transportation industry would be right. 6 So putting that in this context I 7 certainly think that any time there is an adverse 8 consequence to an individual that we get to the 9 point where there is room for a judicial 10 intervention, an independent intervention. 11 That's why I sort of like what the 12 President has done in adding the reasonable 13 articulable suspicion trigger to the querying of 14 the 215 database because that's the point at 15 which some individual becomes, you know, out of 16 the mass pulled out for individuated scrutiny, 17 and that's the point at which he begins to at 18 least suffer the risk of inaccurate adverse 19 consequences. And so I sort of like that as a 20 transition point. 21 MS. WALD: Good. 22 Mr. Felten, you talked about the

- tendency for institutions, including the
- government, to build ever larger databases and
- then to aggregate them. And I think you've said
- 4 either today or in some of your writings that
- 5 there are inherent risks to privacy interests
- 6 when the databases get larger and larger, and
- ⁷ especially when they aggregate them.
- 8 So I guess my basic question to you is,
- 9 what are the principles, and we'll all take notes
- on this, what are the principles that you
- 11 recommend as a computer expert for protecting
- 12 privacy in the increasing use of technology in
- 13 this field?
- I mean all along the way from
- collection, aggregation, whatever you think, if
- somebody wanted to design a system in broad
- 17 concepts that maximize privacy but took adequate
- concern for security, what would they look to?
- MR. FELTEN: Sure. Well, I think the
- first principle would be to try to look beyond
- the most brute force technological approach,
- which is collect all the data that might turn out

- to be useful and retain it all in a single large
- data center for as long as you can.
- The more data you have, the more you
- 4 collect, the greater the adverse consequences
- 5 could be, the greater the risk and the more of a
- target it is, either for abuse or for breach.
- ⁷ So the first principle is to try to fit
- 8 the practices to the specific, to think in terms
- 9 of what kind of analysis it is that you know you
- need to do and figure out which data you can
- 11 collect and how you can structure a system in a
- way that can do that analysis, while collecting
- 13 less data, holding the data more separately and
- 14 pre-processing or minimizing the data first.
- And there's a growing array of technical
- methods that can do this. And unfortunately,
- this becomes a technical problem.
- So the key principle here is simply to
- insist that that technical work be done to try to
- 20 architect a system to collect and hold a minimum
- 21 of data.
- MS. WALD: Who should do that, the

- ¹ government or private industry?
- MR. FELTEN: In my view if, say, a
- 3 government agency wants to argue that they have a
- 4 need to collect and use certain data, there
- should be some onus on them to justify the
- technical practices they're using to justify the
- amount of data collected, the way they're
- 8 organizing it and so on.
- 9 That those who would argue for a
- collection and use of data should be prepared to
- 11 discuss these issues and offer a technical
- ¹² justification.
- When it comes to private parties that's
- a more complicated discussion. I think that the
- best practice in industry ought to be to do that
- as well, although obviously the legal and market
- mechanisms that drive that relationship are very
- different.
- MS. WALD: My last question I'm going to
- throw out, and you can all take a whack of it if
- you want to, but several of you, and I think you
- especially, Ms. Goitein, have talked about the

- element of control of information as being
- essential, but then some other people have
- written in the field said, well, that certainly
- 4 can't be an absolute value. There's got to be
- ⁵ balance. I mean we wouldn't be able to have the
- 6 kind of national security programs if indeed
- everybody said, well, I'm keeping control over
- 8 that piece of information because I don't want
- ⁹ anybody to have it, etcetera.
- So what kinds of principles do you
- 11 apply? Because I assume you recognize that some
- balancing, even as Paul pointed out that even in
- the Fourth Amendment there's an unreasonable
- clause which gives you a kind of a balancing to
- talk about, so how would you handle that?
- And then everybody can take a whack at
- it, and then my panelists will take some more
- ¹⁸ whacks.
- MS. GOITEIN: So as I said before, I
- think in the vast majority of circumstances, and
- that's the way it should be anyway, the drafters
- of the Fourth Amendment did that balancing for us

- and gave us what the government had to do to
- override the privacy right, and that is to show
- 3 probable cause of criminal activity.
- 4 There are some very narrow exceptions
- that the Supreme Court has recognized, some of
- 6 which are controversial, some of which are not.
- And again, so we're not starting from
- 8 scratch. We have to follow the Supreme Court
- go case law here. We can't just say, well, I think
- this particular search was reasonable, even
- though there wasn't a warrant. If it doesn't
- 12 fall within one of the delineated exceptions for
- the court, you have to get a warrant based on
- 14 probable cause.
- Within those exceptions there is for
- balancing, and that's part of the reasonableness
- 17 analysis. What I would say about that is, first
- of all, the courts do their balancing when they
- do a review.
- But Congress has a role as well. And
- when Congress does the balancing on behalf of the
- people, I would agree with what I believe Dan

- said, which is that this is a choice for the
- public to make. This needs to be a public choice
- and it needs to be an informed choice, not a
- 4 choice that's made in secret by a small number of
- officials, but by the public, because this is a
- 6 democracy.
- So we need to have the information about
- ⁸ what the security is, how that threat could be
- ⁹ mitigated by the collection of this information,
- and what exactly is going to be the effect on
- 11 either side.
- The other quick point I would make is
- that in balancing tests national security is too
- 14 often a trump card. The words are uttered and
- we're done.
- And Julian Sanchez from the Cato
- 17 Institute made an excellent point, which is when
- 18 you look at how courts weigh national security
- 19 against the individual interests in question,
- they tend to weigh national security writ large
- over that person's particular interest in that
- information. And that's not the right

- 1 comparison. You either need to weigh that
- person's particular interest in that particular
- information against the incremental threat to
- ⁴ national security in that case, or you need to
- 5 address national security writ large, weigh
- 6 national security writ large against the values
- ⁷ that privacy serves in our society.
- And when you think of it that way,
- 9 national security really shouldn't be a trump
- 10 card.
- 11 You know, we talk about these values as
- being in competition. I think the evidence for
- the most part shows that targeted surveillance is
- more effective than dragnet surveillance. But
- when they are in conflict there needs to be a
- 16 fair and public balancing.
- MS. WALD: Okay, thank you. Some other
- 18 comments? I'll let everybody have a shot at this
- so we can go down the line. We'll start at that
- end with you.
- MR. FELTEN: Okay, thanks. I think in
- thinking about these issues of control, it's

- important to recognize the ways in which people
- try to reassert control, even if they don't have
- ³ it legally.
- 4 And I'm referring specifically to
- technical self-help measures that people use to
- try to limit the flows of information, to try to
- obfuscate identify and behavior, as well as
- 8 strategic behavior in which people avoid doing
- ⁹ certain things or deliberately do certain things
- in order to present a different kind of image to
- whoever it is that they worry is looking at their
- 12 data.
- And these things have substantial costs.
- And I think if you're trying to do a kind of
- utilitarian balancing like Paul was talking
- about, you need to take into account the ways in
- which resources are spent and sometimes are
- 18 really wasted in a kind of arms race between
- self-help and strategic behavior on the one hand,
- and attempts to overcome that on the other side.
- 21 And those costs can often be
- 22 substantial. Just ask any teenager about their

- online use and what you'll hear, and privacy, and
- what you'll hear is an elaborate story about
- 3 technical countermeasures and strategic behavior.
- 4 MS. WALD: Paul?
- MR. ROSENZWEIG: I think your point is
- ⁶ generally well taken, which is to say that
- ⁷ fundamentally the notion of control is at odds
- 8 with government collection of information,
- ⁹ whether it's for the purpose of imposing a tax
- under the IRS, or law enforcement, or national
- 11 security.
- 12 That doesn't mean that it's not an
- important value. It is one that many would
- 14 advance, and I see no reason to discount that at
- 15 all.
- But in some ways if you advance that as
- the touchstone of what you mean by privacy,
- 18 you're setting privacy ineluctably in opposition
- 19 to effective government action in a host of areas
- 20 where people might reasonably want to control.
- You know, I'm sitting here as a
- Republican on the panel thinking of all the

- friends I have who are Second Amendment people
- who think that the government should not collect
- any information about their gun ownership. And,
- 4 you know, that's a perfectly reasonable position
- for them to have. It's not one that we currently
- 6 accept as society.
- And then the last point I would make,
- which is just in response to Liza, because she's
- 9 mentioned it twice, but when I was last in
- government the percentage of searches that were
- 11 conducted without warrants was actually quite
- high, on the order of 50 percent.
- Now I don't know if that's changed much
- because it's been a while since I've been a
- prosecutor, but many, if not most of our typical
- interactions with law enforcement are adjudicated
- on an ex-post reasonableness standard rather than
- an ex-ante warrant standard.
- I don't actually have the data so I
- 20 can't say more about it than that, but I
- 21 certainly seem to recall that it's not always a
- pre, as opposed to a post activity, judicial

- ¹ review.
- MS. WALD: Dan, you have the last word.
- MR. SOLOVE: Sure. A few really quick
- 4 points. First of all, even if you can't always
- ⁵ give people a total control, there are certain
- 6 partial things that we can give people for
- ⁷ control.
- And the other thing is that it's not
- ⁹ just people being in control, it's that the uses
- and gathering of the information is under
- 11 control. And that's another important thing
- about it, that there's appropriate oversight and
- 13 accountability and controls on that gathering
- ¹⁴ too.
- On the Fourth Amendment, I think that it
- would be wrong just to track existing Supreme
- 17 Court interpretations of the Fourth Amendment,
- which I think are a lot of times flawed in a lot
- of cases.
- In fact, I think there are a lot of
- exceptions to the warrant requirement, a lot of
- instances where the Fourth Amendment doesn't even

- get applied at all because the court has this
- ² platonic conception of privacy that is incredibly
- 3 narrow.
- 4 And that's how we get the third-party
- doctrine and how we get a lot of bodies of Fourth
- 6 Amendment law that often take the Fourth
- ⁷ Amendment away from any kind of approach.
- Now the Fourth Amendment, I think, is a
- ⁹ utilitarian balancing. It says basically the
- right to be secure against unreasonable searches
- 11 and seizures. It actually doesn't say privacy,
- it says actually a right to be secure against
- unreasonable searches and seizures.
- And I think that means that any time the
- 15 government is engaging in searches, and
- surveillance, and gathering information that it
- is unreasonable if it's creating problems that
- are not adequately dealt with the right amount of
- oversight and accountability.
- 20 And that's really what the Fourth
- 21 Amendment is trying to impose there, either
- justification to gather information, such as a

- warrant and probable cause, or appropriate
- oversight to make sure that an independent
- judicial body looks at what the government wants
- ⁴ to do and evaluates it.
- I think it's very important that we
- 6 conduct the balance between privacy and security
- ⁷ appropriately.
- I'm not a privacy absolutist, I think
- ⁹ that there should be a balance. But I think it's
- very important that when we balance, we balance
- it correctly and not incorrectly, and that we
- don't skew the balance too much to the security
- 13 side by overweighing the security interests,
- because it's not the entire security interests on
- 15 the scale.
- 16 It's the marginal difference between a
- 17 security interest without certain kinds of
- oversight and accountability, and the security
- interests with oversight and accountability.
- 20 And I think all branches should be
- 21 playing a role and have a role to play in this.
- 22 Congress in the 1970s at the Church Committee,

- which did an extensive review of intelligence
- ² agencies, produced a very illuminative public
- ³ report about that.
- 4 Congress hasn't done anything quite like
- ⁵ that since. I think it should.
- I think the judiciary has a role to
- ⁷ play. I think this body has a role to play. I
- 8 think the people ultimately also are the key to
- ⁹ all this. They have a role to play.
- MS. WALD: Thank you. We're now going
- to have 20 minutes of questioning from my fellow
- Board members, and I think I'll start with the
- 13 Chair.
- MR. MEDINE: Great, thank you.
- Liza raised a question about the proper
- 16 standard for privacy and referenced the Katz
- decision essentially, on expectation of privacy
- and how in some ways people rely on practical
- obscurity because the government is too complex
- or burdensome to gather information.
- In some ways in the computer age we're
- beyond that, which is that the court file that

- was gathering dust now is easily accessible and
- ² public.
- The question is, how should we look at
- 4 privacy issues when public databases are so
- ⁵ readily available?
- And there's also a reference to the fact
- ⁷ that the line between government and commercial
- 8 databases isn't always great and the government
- 9 can access commercial databases.
- So how do we look at privacy when the
- information is out there, it is publicly
- available, but yet, as Ed pointed out, you
- 13 combine it into a mosaic and it can create a very
- detailed profile, and should the government be
- collecting that information?
- So what standard should we apply in this
- 17 context? What's the Katz 2014 version as far as
- 18 how the government ought to recognize privacy
- 19 issues?
- And I'm happy just to go down the line.
- MS. WALD: Well, whoever wants to take
- it. I might note that time-wise, we're going to

- have about five minutes per, so if you can keep
- your answers or your comments relatively brief,
- we can make sure that everybody gets their full
- 4 component of time.
- MR. SOLOVE: I'll be super-brief. I
- think that right now what's been known as the
- mosaic theory that we see in the concurrences to
- 8 the Jones case in the Supreme Court are starting
- ⁹ to look at this very question.
- I can't really answer it in a few
- seconds, but I think it's to look at when we
- combine various pieces of data, what are the
- implications of that and when does the combining
- of that data reveal new information that can
- 15 create certain problems and harms to people, and
- that's where we want to step in.
- MR. ROSENZWEIG: I'd make two quick
- 18 points. The first is of course that practical
- obscurity is itself a sort of a post-industrial
- concept.
- 21 If you were in a medieval village back
- in the 1200s there was no practical obscurity.

- You were limited to who you knew and they knew
- everything about you, pretty much. The data
- 3 aggregation system was the coffee klatch where
- everybody talked to each other.
- 5 So in advancing practical obscurity
- 6 we're advancing a value that has come to be
- ⁷ something that we value more now, one that I
- ⁸ agree with.
- I think Dan's exactly right, the mosaic
- is real. To deny that is to deny the reality of
- the science that Ed knows.
- So it strikes me that the most likely
- 13 points of intervention are either at the
- collection, or at the aggregation, or at the use
- of the aggregated data.
- I tend to think you can't do it at
- 17 collection because the databases are there.
- They're so big it's impossible to stop. Unless
- 19 you're going to stop Google from collecting,
- we're going to have big data collection.
- So it's got to be when the government
- chooses to aggregate it or perhaps chooses to act

- upon the aggregation. And as between those, I
- don't have too much choice.
- MS. GOITEIN: Just a quick note, I agree
- 4 with most of what's been said in terms of the
- 5 mosaic theory.
- I mean another way to look at it is just
- ⁷ that the information that's being gathered by the
- 8 government is, in fact, information that using
- 9 normal powers of human observation would be in a
- person's control and would not be something the
- 11 government would have access to.
- The one thing I would say is that I
- don't agree that the point of collection is a
- moot point because the mere fact that Google has
- all of this information and Facebook has all of
- this information does not mean that the
- government has all of this information.
- And there are burgeoning new
- technologies that their use has not been decided,
- such as UAVs and how the government will be
- 21 allowed to deploy UAVs. So there's still plenty
- of room to regulate at the collection phase.

- And for all the reasons we discussed
- earlier about chilling effect and about what
- privacy means to different people, I think that
- is the point at which the privacy interests
- ⁵ arise.
- 6 MR. FELTEN: And I'll be very brief as
- ⁷ well. Along with what the other panelists have
- 8 said, I'd also point out that much of the
- 9 information that is in corporate databases is
- information that was observed rather than
- disclosed, and there's not always consent, or at
- least often consent is very thin from the person
- who the data is about.
- And so I don't think you can always
- infer that there was an awareness. You can't
- infer from the fact that information is in, say,
- a corporate database that a user was aware that
- it was collected or that they were aware that it
- might go to the government and be used for
- government purposes.
- MR. MEDINE: And therefore should the
- government not collect the information under

- 1 those circumstances?
- MR. FELTEN: Well, I hesitate to make a
- legal opinion here, not being a lawyer.
- 4 MR. MEDINE: As a policy matter.
- MR. FELTEN: As a policy matter?
- 6 MR. MEDINE: Yes.
- 7 MR. FELTEN: But I should say that as a
- 8 policy matter I get very nervous when it appears
- ⁹ that there is a legal fiction that something has
- happened when it's clearly not happening. So a
- 11 fiction of consent or a fiction that the mosaic
- effect doesn't exist are troubling.
- MR. MEDINE: My time has expired.
- MS. WALD: Rachel Brand.
- MS. BRAND: Thank you. Thank you all
- ¹⁶ for being here, first of all.
- Going back to this notion of control
- that Judge Wald was asking about, Ms. Goitein,
- 19 you went to the Fourth Amendment concept. I'm
- interested in whether the notion of control
- that's embodied in the FIPPs, which is more of an
- individual participation concept, can apply in

- the national security surveillance context.
- So I think one of you noted that maybe
- no individual would say, yes, I consent to being
- surveilled by the NSA, or the FBI, or anybody
- ⁵ else, and if that were the standard then you
- 6 couldn't have surveillance programs.
- And the FIPPs is on top obviously of
- 8 whatever the Fourth Amendment baseline is. FIPPs
- 9 would impose on government agencies additional
- 10 restrictions.
- 11 Can that kind of notion apply at all in
- the national security context? What's your view
- on that?
- MS. GOITEIN: I think it can apply but
- 15 I'm just sort of pausing because I'm thinking
- about some of the premises of the questions.
- 17 It's not the case that you couldn't have
- surveillance programs if people didn't consent to
- the disclosure of their information. The
- government can obtain your information with a
- warrant based on probable cause.
- MS. BRAND: No, no, my point is that

- $^{
 m 1}$ we're beyond the Fourth Amendment now. We're
- layering on top of the Fourth Amendment the FIPPs
- 3 individual participation notion.
- 4 And the reason I ask is because, for
- ⁵ example, when the NSA published their report on
- targeted data collection under 12333, they said
- ⁷ that they were applying the FIPPs, but then they
- 8 turned right around and said, but the individual
- 9 participation concept doesn't apply so we're not
- applying that part of it.
- So what I'm wondering is whether the
- 12 FIPPs is just not the right, I don't know,
- 13 framework to apply in this context. That's what
- 14 I'm trying to get at.
- 15 Is it, does this individual
- participation thing just not apply and should we
- look for some other framework or standard to use?
- 18 That's what I'm getting at.
- MS. GOITEIN: Actually if you wouldn't
- mind I'd like to think about the question.
- MS. BRAND: Okay.
- MS. GOITEIN: I have some thoughts but I

- want to think about it a little more and maybe I
- ² can put it in writing along with my testimony.
- MS. BRAND: Okay.
- MR. SOLOVE: I have a thought on it. I
- think the FIPPs model has some flaws to it. You
- 6 know, a lot of times people don't read the
- 7 privacy policies, in most cases, of companies.
- 8 And I'm not sure just providing a notice is
- ⁹ effective.
- So we do need to think about what works
- in this context. I think that the key is in
- certain cases we might want individuals to play a
- greater role. I think the TSA, if you're on the
- 14 no-fly list, I think you should have a right to
- be heard. There should be rights of redress
- there and to challenge your being on that list.
- So I think there some of the FIPPs make
- 18 a lot of sense. Some of the FIPPs like security
- 19 I think make sense. Other ones might not.
- But I think the larger component of all
- this is that there's adequate control and
- accountability, which is also part of the FIPPs.

- So while everything in the FIPPs, such
- ² as an individualized notice of every time
- information is collected is not really feasible,
- ⁴ there are certain things.
- 5 There's a greater transparency right in
- the FIPPs too, not that individuals get notified
- about every collection about them, but that
- 8 there's a public accountability and a generalized
- 9 disclosure about what's going on.
- MR. ROSENZWEIG: I thought that the
- 11 acknowledgement in the NSA report that some of
- the FIPPs principles simply could not be fully
- 13 implemented in the context of a national security
- surveillance program was an absolutely accurate
- acknowledgement of reality.
- You can't provide error collection
- notice in all circumstances. I certainly agree.
- 18 I mean was talking more about the secondary
- screening than the no-fly list where we do have
- more robust rights.
- But the challenge for you is going to be
- trying to figure out what the underlying values

- 1 are and how to get at those.
- So in this context I think the
- ³ underlying value is prevention of governmental
- 4 abuse, that's what animates everybody in this
- ⁵ sphere, and government surveillance modifying
- 6 behavior.
- And the types of accountability and
- 8 transparency that you have to help build are ones
- ⁹ that match the operational needs of the national
- security system, while providing protections
- 11 against that.
- We tried that with the intelligence
- committees and the post-Church Commission
- 14 modifications, something that we might call kind
- of delegated transparency where we all trust the
- 16 Congress to do it right.
- 17 It seems as though we're less willing to
- do that now. Personally I'm not so certain that
- that's a good impulse, but it seems like that.
- So maybe it's this Board. Maybe it's a
- judicial panel with a cleared advocate in front
- ²² of it.

1 There are lots of mechanisms, short of 2 the complete transparency and accountability and 3 individual participation that are part of FIPPs that could be imagined that would achieve the 5 objective of controlling against governmental 6 abuse and misuse, while not completely frustrating the operational necessities that I think most of us see as remaining regnant. 9 So I think a lot of it would be things 10 that are more in Ed's bailiwick, which are 11 thinking about what the use case scenarios that 12 are legitimate are in advance and building in 13 enhanced privacy protections on a technological 14 level. And then you can have as much of your 15 cake as you want and still get to eat some of it. 16 MR. FELTEN: To the extent that 17 particular principles from the FIPPs might be 18 difficult or impossible to apply in this kind of 19 setting, it seems there should be a greater 20 obligation to further the goals of that 21 principle. 22 So, for example, if you can't offer the

- 1 right to control or correct errors in the data,
- you could imagine asking for greater effort to
- ensure the correctness of the data as it is, or
- 4 extra safeguards ex-post regarding the
- ⁵ possibility of error.
- 6 MS. BRAND: Thank you.
- MS. WALD: Okay. Did you collect any
- 8 thoughts that you, very briefly --
- 9 MS. GOITEIN: Yes, and I think I would
- agree with Ed. I mean part of what I was
- struggling with is how much are we giving up on
- this sort of collection, which I'm not quite
- willing to do, in talking about sort of, you
- know, post-collection? And that's why I wanted
- to go back to that issue of surveillance and
- control over the information.
- I still want to go back and look more.
- 18 This is honestly something I just haven't thought
- about enough and so I do want to go back and, you
- 20 know, look at FIPPs, which I used to use all the
- time when I was on the Hill to craft our privacy
- amendments, but I want to go back and look more

- ¹ carefully.
- But, you know, it sounds to me like the
- best approach is a modified version of the FIPPs,
- ⁴ but I want to look more closely.
- MS. WALD: That's fine. We'll be glad
- to receive any later submissions from any of the
- 7 panelists.
- Before we go on to Beth Cook's questions
- ⁹ I want to remind the audience that if you have
- any questions, write them down and they'll bring
- them up to me and then I will -- okay, they're
- coming, that's good to know.
- Okay, Ms. Cook.
- MS. COLLINS COOK: So thank you all for
- what I've found to be a very, very interesting
- panel. I hope it bodes well for the rest of the
- ¹⁷ day.
- And in fact, a lot of, I think panel 3
- will be dealing exactly with how you translate
- the FIPPs, is that the right transition, does
- that really work in the government context.
- But I was also struck by the numerous

- 1 mentions of mosaic theory because there are
- obviously other implications of the mosaic
- 3 theory.
- One bears on transparency, which is to
- the extent that we are transparent in seemingly
- discreet ways, our adversaries are also looking
- ⁷ to aggregate information about sources and
- 8 methods.
- The other is that the national security
- establishment I think would argue that the mosaic
- theory is critical. You need to understand
- mosaic theory to understand collection, to
- understand exactly how the national security
- apparatus works, that they have to be able to
- ¹⁵ aggregate information.
- You can agree or disagree, but I think I
- was struck by the different implications of the
- mosaic theory.
- So I wanted to start with you, Professor
- Felten, and I was really interested in your
- notion of moving away from the brute force
- 22 collection mechanism.

- And I think the Section 215 program is
- one that the government had made the argument
- 3 essentially that they need the brute force
- 4 collection, they need to have the retention in
- order to identify previously unknown links and
- 6 information.
- ⁷ Have you given thought as to whether or
- 8 not there are technological options available to
- 9 limit collection for a program like Section 215?
- If you haven't, then more generally if
- could you be more specific about collection
- options.
- MR. FELTEN: Yes. Well, with respect to
- 14 Section 215, the data of course is collected
- initially by the phone companies, right? And
- there's the question of whether the information
- needs to be transferred in bulk to the
- intelligence community in order for them to be
- ¹⁹ able to do their analysis.
- 20 And I think it's pretty clear that as a
- technical matter the kinds of linking, looking
- for, say, multi-hop links that the intelligence

- agencies want to do, can be done technically
- while the information is still held by third
- 3 parties such as the phone companies.
- 4 This requires a modest amount of
- technical coordination between the companies, the
- 6 entities holding the data and the entities that
- ⁷ are doing the analysis.
- So there are opportunities to match, to
- 9 look for whether there are paths of two hops or
- three hops from point A or point B, etcetera, and
- then to reach in and extract just the data items
- of those individuals or phone numbers that are
- highlighted by that kind of analysis. That's the
- 14 kind of thing that can be done.
- There's further work that is more
- technical that goes to questions of whether you
- can use, say, advanced cryptography to allow that
- same analysis, while not disclosing to the phone
- 19 company information about which numbers are being
- searched or linked.
- 21 And those sorts of methods are, I'd say
- developing, and there's been some interest in the

- technical problem of how to do this in the
- independent research community in light of what
- we've learned publicly about the Section 215
- ⁴ program.
- 5 And one of the lessons of that is that
- 6 methods are often available or developable when
- you have a specific technical problem like this.
- MS. COLLINS COOK: I think our biggest
- 9 challenge is taking the concepts that we're
- talking about today and developing practical,
- 11 feasible recommendations that can actually be
- 12 implemented.
- So the more concrete and the more
- specific that we can be in terms of
- 15 recommendation, the more likely they are to be
- implemented.
- Briefly, both to the professors in the
- middle I would ask, you both talk a little bit
- about risk mitigation, and assuming that there
- are going to be harms, how do you mitigate those
- harms past the collection stage?
- What have you found to be the most

- 1 effective mechanisms for mitigating risk? Is it
- ² retention periods? Is it access controls? Is it
- 3 audit trails? So what can the government do
- 4 concretely to start mitigating risks?
- MR. SOLOVE: Well, I think it's not
- for really just one thing that I can sort of point to
- ⁷ like, that is it. It's all of those things are
- 8 very valuable to do, everything from mechanisms
- ⁹ to ensure that this information is accurate when
- information is grabbed from one context to
- another.
- You know, what's accurate enough for the
- purposes of Amazon.com to recommend books for you
- 14 is not the same level of accuracy we might want
- 15 from the government. So if Amazon makes a
- mistake and recommends the wrong book to you, big
- deal. It doesn't need a hundred percent accuracy
- ¹⁸ for that.
- But the level of accuracy differs as we
- differ in context. So we need to have mechanisms
- to make sure that when information might be taken
- from one context and put into the other that it's

- appropriately accurate for that particular
- ² context.
- We need an analysis of how long we keep
- 4 the data, audit trails to make sure that it's not
- ⁵ being improperly accessed, appropriate
- 6 accountability to make sure it's being kept
- ⁷ adequately secure, and also how it's being used,
- 8 controls on its use so it can't be used for any
- 9 purpose ten years from now.
- So we need all these different things,
- and oversight from a lot of different bodies, I
- think. So it's actually a complex thing with
- many, many parts.
- MR. ROSENZWEIG: There are certainly
- many moving parts, but from my perspective both
- 16 from outside and when I was inside, since the
- threat that we're talking about is governmental
- abuse or misuse is the primary one, the principal
- 19 factors that I would focus on that seem to have
- been effective were ones that focus on the
- 21 individual government actors.
- Training in the first instance so that

- they know the rules, inculcating a culture of
- ² compliance that is pre-error mechanisms, then
- obviously a lot of audit compliance work from
- 4 outside inspectors general and/or Congress.
- 5 And then finally, and this is perhaps
- 6 where we fall down the most, the willingness to
- ⁷ impose at least administrative sanctions on
- 8 people who vary from the accepted rules, at least
- ⁹ in a willful context, and even perhaps in a
- 10 negligent context.
- You know, nothing attracts the attention
- of a government employee so much as the prospect
- of losing his job or being suspended for a term
- 14 of months. So that would be where I would focus.
- MR. FELTEN: If we look at the failures
- of compliance that have been acknowledged, we see
- some of them that are individual employees doing
- things they shouldn't, but we've also seen some
- 19 that are failures of the technical systems to
- behave consistently with the internal policies.
- 21 And this is a case where oversight can
- operate without needing to get deeply into the

- 1 nuts and bolts of the technology, just the
- question of what processes are in place to make
- 3 sure that your technology does what your general
- 4 counsel says it should do. And I think there's
- 5 an opportunity to push on oversight in that area.
- MS. WALD: I think we'll move to Jim
- 7 Dempsey now.
- MR. DEMPSEY: Thank you, and thank you
- ⁹ to all the witnesses. I think it's very
- important as we wrap up this panel to highlight
- what I at least heard is an awful lot of
- 12 commonality.
- Because I think that it's important to
- the Board and important for the public debate
- moving forward not to end up with the proposition
- that this is all so confusing, or this is all
- disparate, there are so many different views.
- I heard actually a lot of commonality
- among the witnesses, starting with the point that
- I think you all agree, whether you start from the
- 21 premise that privacy is a human right or whether
- you start from the premise that it's an

- instrumental right, I think all of you agree that
- it's an umbrella term which covers many different
- yalues, many different interests.
- 4 And I also heard agreement that the
- mosaic theory, even if it hasn't been accepted by
- 6 courts, is real. It's real, both from a privacy
- perspective and it's real from the governmental
- 8 perspective.
- 9 MS. GOITEIN: Let the record reflect
- 10 nodding.
- MR. DEMPSEY: And thirdly, I think I
- 12 heard unanimity that what the law refers to at
- least as the third-party doctrine, the doctrine
- that by giving information to one person you lose
- all interest, all privacy interest in that
- information, that disclosure to one surrenders
- your right with respect to disclosure for any
- other purpose, again, am I right there was
- 19 agreement that that concept of disclosure to one
- is disclosure to all is not valid,
- 21 constitutionality aside, for modern day reality,
- that doctrine just doesn't fit with the way we

- view information and the way we view privacy?
- And Dan is nodding. Paul, would you
- 3 agree that disclosure to one is not a surrender
- 4 of all interest in the information?
- MR. ROSENZWEIG: I would say that the
- 6 way that people interact today it would be
- ⁷ inappropriate to imply consent to universal
- 8 disclosure from explicit consent to disclosure to
- ⁹ a single person, yes.
- I'm not sure that I would agree with
- what's implicit in your question, which is that
- it necessarily follows that that is a matter of
- either constitutional significance or one of
- legal cognizable significance that should animate
- this Board. I want to think about that.
- But I would certainly accept the premise
- 17 that human experience is that if I tell Dan a
- 18 secret, I'm not expecting him to tell everybody.
- MR. DEMPSEY: In fact, there's an
- instrumental approach, there's an instrumental
- value that the disclosure of your medical records
- to the doctor is specifically premised on the

- notion that they are, thereby you have not
- surrendered your privacy rights. And in fact, we
- want people to accurately disclose information to
- 4 their doctors, therefore we promise them that
- their medical records, disclosure to the doctor
- 6 is not disclosure to all.
- 7 MR. ROSENZWEIG: That's true, though of
- 8 course that's a wonderful example because we
- 9 accept statements made to a doctor as an
- exception from the hearsay rule precisely because
- we think that when you talk to a doctor in an
- emergency situation you're motivated to actually
- 13 be telling him the truth. I was shot, doc. And
- so the doctor can in some circumstances be
- compelled to. So those realities work both ways.
- MR. DEMPSEY: Can be compelled but not
- 17 obviously --
- MR. ROSENZWEIG: Not obviously
- 19 collected. Not collected under, yes, HIPAA.
- MR. DEMPSEY: Right, yes.
- 21 Also there were several witnesses
- mentioned the FIPPs. And I think it's, first of

- all, important to say we're talking about the
- Fair Information Practice Principles, which
- ³ actually there's no definitive version of them.
- But there is a version that was adopted
- by the Department of Homeland Security in 2008,
- 6 which is as good as any, I think.
- And it seemed to me also that there was
- 8 agreement that they are, the FIPPs framework
- ⁹ provides the framework, the questions.
- They're nowhere perfectly implemented,
- they're nowhere fully implemented, but they are
- 12 relevant as a framework for asking about how you
- deal with information.
- And then you decide, do you adjust it,
- does it work? If it doesn't work, do you
- 16 compensate for it with more emphasis on other
- issues? Is that again a fair --
- Paul, you're making a somewhat skeptical
- 19 face, but you at least can say that it is a
- 20 framework for asking the questions.
- MR. ROSENZWEIG: It's a framework for a
- starting point for asking the questions, but I

- think that many of those questions don't
- withstand the technological transitions we're
- ³ going through.
- 4 And so I accept it as a leaping off
- point, but I think I'm probably more willing than
- 6 some of the other members of the board to discard
- ⁷ some of them as inoperable under current
- 8 circumstances.
- 9 MR. DEMPSEY: And what would you replace
- 10 them with?
- MR. ROSENZWEIG: Well, as Ed said,
- emphasis on the remaining aspects and then, to my
- mind, I think kind of a more granular analysis of
- the underlying interests at stake and thinking
- about what the mechanisms are, the privacy
- interests that we're talking about is that we
- have to protect.
- Because, you know, FIPPs is kind of one
- size fits all, and I just don't think it kind of
- 20 covers the range of the privacy interests that
- the chairman outlined so ably, so ably, earlier
- in the day.

1 MR. DEMPSEY: Okay, thank you. 2 MS. WALD: Okay. We have a couple of 3 questions from the audience. I'm not sure we're going to get to all of them, so what I'm going to 5 do is direct them. I'll just be arbitrary and 6 direct them to a particular panel member, and then if you can keep it as brief as you possibly 8 can. 9 The first one, actually the writer 10 wanted it directed toward you, Ms. Goitein. When 11 a government draws data from private databases, 12 such as telephone metadata, at which point of 13 collection is more regulation required, the 14 private entity's collection or the government 15 collection from the private entity? 16 That's a yes or no. 17 MS. GOITEIN: I was going to say, I 18 don't think I can answer that question. It just 19 depends what you mean by more regulation. 20 I think obviously when you disclose 21 certain information to your telephone company, 22 you are in a contract with that company and that

- 1 contract regulates your dealings with the
- ² company.
- I think one of the problems with the
- 4 metadata program is that there was no reading of
- ⁵ either the contract or Section 215 of the PATRIOT
- 6 Act that would have enabled any person to know
- yhat they were consenting to and to know that
- 8 their information would then go to the NSA.
- 9 MS. WALD: Your answer is both?
- MS. GOITEIN: It's both. There's just
- different types of regulation. There's the
- 12 contractual regulation.
- There is some degree, I mean the Stored
- 14 Communications Act is government regulation, when
- you get certain kinds of information from the
- telephone company.
- And then for the government there's the
- 18 Fourth Amendment. And there's all manner of
- laws, so lots of regulation everywhere. I know
- ²⁰ that's --
- MS. WALD: Okay, for you, Dan. I think
- 22 it was Liza Goitein that said that private

- companies have no incentive to coerce or imprison
- people, that's why perhaps the risks of injury
- might be greater from the government than from
- ⁴ private companies.
- But the writer asks, does that take into
- 6 account the homeland security and prison
- industries? NSA couldn't do what it does without
- ⁸ 484 contractors providing IT technical support.
- ⁹ Are there risks inherent in the increasing
- commercialization of national security?
- MR. SOLOVE: Well, yes, I definitely
- think problems can come from anywhere, and I
- don't think there's sort of inherent things that
- can be said about, you know, various things about
- where problems could be caused.
- I think we want to look at, you know,
- when does the collection and the amassing of data
- by the private sector cause problems? When does
- that access by the government create problems?
- And increasingly we see a cooperation or
- 21 an industry in the private sector that has grown
- up to basically perform government functions and

- 1 help gather data, help analyze data and then
- share data with the government.
- I think all these things create various
- 4 problems that we need to address. And so I think
- if we both keep our eye on the problems and stop
- 6 looking elsewhere and just look at the problems,
- and we address those problems wherever they may
- 8 happen, I think that's the best approach.
- 9 MS. WALD: Okay. Here are two, I think
- this must go to you, Ed Felten, could the
- 11 panelists discuss what they think their Tesla, I
- 12 had to ask what that was, of today should
- 13 provide, what technologies of data flow analysis
- 14 could or should be built in?
- I know you've covered a great deal of
- this before, so if you could just give us a one
- or two sentence summary on it, that would be
- 18 fine.
- MR. FELTEN: In a sense the question is
- asking me to sum up sort of a whole area of
- 21 knowledge in a few seconds, which I won't try to
- ²² do.

- I'd simply say that as with cars, as
- with the Tesla, you know, some sort of high end
- 3 car, you should think in terms of which
- 4 technologies are available and reasonably
- ⁵ practical to use to minimize, or control, or
- 6 limit the risk of certain information practice,
- ⁷ and then ask that those be there.
- 8 You should ask that an entity that wants
- ⁹ to collect and use the information be willing to
- justify the choices they've made and be willing
- to justify why they did not use some accepted
- technical, privacy-preserving technical method if
- it seems to be available.
- MS. WALD: Okay. The last one is, Paul,
- 15 I don't think this is in your natural bailiwick,
- but I'll pick you anyway.
- MR. ROSENZWEIG: Okay.
- MS. WALD: What about the application of
- 19 privacy in quasi-federal organizations like the
- 20 Postal Service or the PBGC?
- 21 If I can remember back to my old
- judicial background, that's something benefits

- 1 quarantee corporation.
- MR. MEDINE: Pension benefits.
- MS. WALD: Pension Benefits Guarantee
- 4 Corporation. How are they impacted by the Fourth
- 5 Amendment? Are there issues and concerns for
- 6 privacy in those organizations?
- 7 MR. ROSENZWEIG: I suppose the honest
- answer would be I'm not sure. But my
- ⁹ understanding is that the Fourth Amendment
- applies to those institutions insofar as they are
- exercising governmental authority and acting as
- 12 agents for the government.
- So I assume that Postal Service
- employees can't open your mail willy-nilly just
- because they're pseudo-private actors. I may be
- wrong about that, but since they don't open my
- 17 mail.
- Jim's nodding, no, I'm right. So
- thanks, that's good.
- I think that the implication of the
- question, which is really the most interesting
- part of it, so I'll transition into something I

- 1 do want to talk about, is that it emphasizes the
- point that Liza made, with which I do agree,
- which is that the line between commercial
- 4 collection and government collection is
- increasingly blurring some, you know, and the
- idea that regulation of the government but no
- 7 regulation of Google's collection kind of sits in
- 8 dissonance. And there are these places that are
- 9 halfway between.
- For me, you know, that suggests one set
- of answers, because I'm unwilling to think about
- wholesale government regulation at an extreme
- 13 level of corporate business practices. I think
- there's some there, but it certainly emphasizes
- the confluence between them.
- MS. WALD: Okay. Well, that ends my
- part of the panel, unless the Chair has some
- 18 parting words.
- MR. MEDINE: Thank you very much.
- MS. WALD: You've been extremely
- 21 forthcoming.
- MR. MEDINE: Thanks to the panel and for

- the audience questions.
- We'll take a 10 or 15 minute break and
- we'll resume at 10:30 with the technology panel.
- 4 (Off the record.)
- MR. MEDINE: Thank you very much. We
- 6 will resume and Jim Dempsey will be moderating
- ⁷ this panel.
- MR. DEMPSEY: Thank you, Mr. Chairman.
- 9 Good morning again to members of the audience,
- particularly good morning to our second panel.
- 11 The title of our panel is Privacy
- 12 Interests in the Counterterrorism Context and the
- 13 Impact of Technology.
- I have no statement of my own, so we can
- go straight to the opening statements by the
- witnesses. I'll introduce each of them in turn.
- We can go down the row, which happens also to be
- 18 alphabetical order.
- 19 I remind the witnesses that we would ask
- them to keep their opening remarks to seven
- minutes. There is a timekeeper, which you might
- not have seen, but in the front row here, Renee,

- who will be holding up a yellow card for your two
- minute warning and then a red card for time's up.
- Thereafter a round of questioning by the
- 4 Board members, and again the possibility of
- ⁵ questions submitted by members of the audience.
- 6 PCLOB staff members throughout the
- ⁷ audience have little index cards, and so if
- 8 during the course of the panel a question occurs
- ⁹ to you, raise your hand and someone will bring
- you over a little 3 by 5.
- Our first speaker or member of this
- panel is Annie Anton. She is a professor in and
- 13 Chair of the School of Interactive Computing at
- 14 Georgia Tech University. She has a Ph.D. in
- computer science, and is one of the country's
- leading experts on issues at the intersection of
- technology and policy.
- So, Annie, please.
- MS. ANTON: Good morning and thanks for
- the opportunity to testify.
- We're in an ever changing world where
- terrorists and criminals are getting smarter and

- 1 more sophisticated. Their offensive techniques
- ² are surpassing our ability to protect our nation.
- Providing strong technical protections for
- 4 privacy and civil liberties is a counterterrorism
- weapon.
- Today I focus primarily on three
- ⁷ technology considerations. First, strong
- 8 encryption is an essential technology for
- ⁹ fighting terrorism.
- Second, de-identification, while not
- perfect, may be a reasonable approach given a
- 12 thorough risk analysis.
- And third, improved privacy threat
- modeling is critical for counterterrorism.
- Our national cyber infrastructure must
- be resilient to attacks from foreign powers,
- terrorists and criminals.
- Requiring government backdoors in
- 19 commercial products, stockpiling zero-day
- 20 exploits and weakening software security
- 21 standards are all practices that weaken our
- nation's cyber security posture and make it

- easier for attackers to infiltrate these systems
- ² for nefarious purposes.
- The latest Apple and Google phones build
- in encryption by default. Both companies are
- 5 configuring this encryption such that they cannot
- 6 decrypt the information for anyone, including law
- ⁷ enforcement.
- These measures have been sharply
- ⁹ criticized by the Director of the FBI and the
- 10 Attorney General.
- 11 As a technologist, I can assert that
- 12 applying security best practices such as
- encryption by default will yield a system that
- can better withstand intrusions and denial of
- service attacks, as well as limit access to
- ¹⁶ authenticated and authorized users.
- 17 Requiring companies provide backdoors
- 18 for law enforcement or national security hurts
- both individual privacy and our nation's overall
- security.
- Moreover, the security benefits are
- questionable at best because sophisticated

- terrorists and criminals will simply use
- international products or more secure, less
- 3 convenient alternatives.
- 4 Technology and policy scholars are
- 5 actively debating the merits of de-identification
- and anonymization techniques. The issue is
- 7 critical because privacy rules only apply to
- 8 identifiable data. Technology scholars emphasize
- ⁹ that there is no way to mathematically prove an
- anonymized data set, that it cannot be
- 11 re-identified.
- In contrast, policy scholars believe
- that anonymization provides real practical
- 14 protection to most of the people most of the
- 15 time.
- 16 Consider that the locks on your door at
- home are pretty good, but not good enough to keep
- 18 a determined intruder at bay. That's the idea
- behind practical anonymization.
- There are some cases where it is
- 21 critical to protect a person's identity. For
- example, for victims of domestic abuse we need to

- ensure that their location is protected and
- ² cannot be re-identified by their abuser.
- However, in many settings, if we apply
- effective but not perfect de-identification
- ⁵ procedures, overall privacy protection may be
- increased and data may be more useful. In such
- 7 cases the perfect should not be the enemy of the
- 8 good.
- The PCLOB might consider how to
- determine in practice when agencies should insist
- on technically strict de-identification versus
- when effective, but not perfect,
- de-identification may address the bulk of the
- 14 risk.
- Finally, threat modeling is critical for
- counterterrorism, and we must improve it to
- 17 achieve two goals.
- First, we must develop privacy oriented
- models. Most threat modeling techniques have
- been developed entirely in a security context
- with little privacy consideration. The latter is
- crucial given the rise the big data analytics and

- ¹ the Internet of things.
- Second, as a nation we do not want
- 3 insiders leaking state secrets to foreign
- 4 journalists to become a common way to influence
- ⁵ public policy decisions and debates.
- Insiders with access to sensitive
- ⁷ information must be considered as potential
- 8 threats simply because of the extreme damage that
- ⁹ a leak could do, either in direct cost by
- 10 providing useful information to enemies, or
- indirect cost with respect to public relations or
- erosion of trust. A good threat model makes risk
- analysis feasible for any organization.
- In closing, as a technologist and
- privacy scholar I believe we should encourage
- strong encryption by default, use practical
- de-identification technologies now rather than
- wait for theoretically perfect solutions, and
- expand threat modeling to include privacy and
- 20 security as well.
- In addition, Ed Felten mentioned the
- importance of having technologists in the room.

- 1 I can't help but note that the review group did
- not have a technologist that the PCLOB, which I
- 3 really appreciate all that you are doing, but
- 4 again, there isn't a technologist in the room.
- 5 And having technologists on panels is
- 6 helpful, but really I would like to see us move
- ⁷ forward to having more technologists actually
- 8 involved in the decision-making.
- And so I'd like to thank the Civil
- 10 Liberties and Privacy Board for its commitment to
- 11 finding ways for the government to protect
- privacy, and also for meeting our critical
- security needs as a nation as well. Thank you.
- MR. MEDINE: Let me just thank you for
- your testimony, but actually we have a
- technologist in the second row.
- MS. ANTON: Great.
- MR. MEDINE: And we have a technologist
- outside as well. And so we actually do value the
- 20 role of having technologists and have two full-
- time on our staff.
- MS. ANTON: Good, and I look forward to

- meeting them. Thank you.
- MR. DEMPSEY: Thank you. Our second
- ³ witness is Alvaro Bedoya. Alvaro is the
- Executive Director of the Center on Privacy,
- ⁵ Technology and the Law at Georgetown University
- 6 Law School and was previously chief counsel to
- ⁷ the Senate Judiciary Subcommittee on Privacy,
- 8 Technology and the Law. Alvaro.
- 9 MR. BEDOYA: Thank you. Good morning
- and thank you for the opportunity to speak with
- 11 you today.
- We have a problem right now in privacy
- and it's a problem for government and industry.
- Government and industry have developed
- extraordinarily powerful data analysis tools.
- These tools let them analyze data sets
- that have previously been too large or too messy,
- they let them process that data faster, and they
- 19 let them find latent value in data sets that have
- 20 previously seemed old and worthless.
- In short, these processes create
- enormous value, and that value is driving both

- 1 government and industry to collect as much
- information as possible and to retain it as long
- 3 as possible.
- The problem is, is that's hitting up
- 5 against long-established privacy values ingrained
- in the FIPPs. The FIPPs encourage limited
- ⁷ collection, they encourage data minimization, and
- 8 the destruction of data that's no longer useful
- ⁹ for the purpose for which it was collected.
- And so right now both in the
- intelligence community and in industry there's
- effectively an effort to redefine privacy.
- Privacy used to be about collecting only
- 14 what you needed to collect. Under the new model,
- you collect as much as information as possible
- and you protect privacy through after the fact
- post-collection use restrictions.
- I'm here to encourage you to resist this
- 19 new model. In my written testimony I argue four
- points. The first is that collection still
- 21 matters. The collection of personal data impacts
- 22 a person's core right to privacy, regardless of

- 1 what happens to that data after the fact.
- Second, this was discussed in the first
- panel, but there's a misconception, I think, that
- 4 the FIPPs are primarily useful for commercial
- ⁵ privacy.
- In my written testimony I talk about the
- ⁷ fact that the FIPPs remain a critical benchmark
- 8 against which to measure the privacy impacts of
- 9 counterterrorism policies.
- And I'll just add given the previous the
- discussion, that literally since their inception
- in 1973, the committee that wrote the report
- dedicated a section, it's just two pages, talking
- about how of course not all of the FIPPs can
- apply in the intelligence context, but clearly
- some of them must because the risk is too high.
- Third, in my testimony I talk about that
- we need to remember that privacy is not about
- taking but about -- pardon me -- it's about
- taking and not about sharing.
- 21 And fourth and finally, I think that
- 22 Americans do expect a degree of privacy in

- ¹ public.
- Now given my limited time here I
- 3 actually want to focus my oral testimony on just
- 4 that first point, collection. I think it's the
- 5 most important.
- 6 After the Snowden disclosures on the
- ⁷ telephone records program last summer, the IC's
- first line of argument was that, you know, we may
- 9 collect a lot of this information but we only
- 10 look at a tiny part of it.
- The problem is that this is not how
- 12 people think about privacy. If a police officer
- 13 knocked on your door and said, hey, I want you to
- give me a list of every person you've spoken with
- in the last week and then said, you know, don't
- worry, we're really probably never going to look
- at this stuff, would that reassure you? I think
- that most people would say no.
- And I think that this highlights the
- 20 fact that the forcible collection of sensitive
- data in and of itself invades what this Board has
- called, "the core concept of information

- 1 privacy". And that's, "the ability of
- individuals to control information about
- 3 themselves".
- It's not just a concept. As you know,
- ⁵ it implicates First Amendment and Fourth
- 6 Amendment interests. I elaborate that in my
- ⁷ written testimony.
- But in my mind the single biggest reason
- ⁹ to resist the privacy model that primarily relies
- on post-collection use restrictions is the
- disparate impact that that model might have on
- vulnerable communities.
- Now again, in a use restriction model
- you collect everything and you protect privacy by
- banning harmful uses of data after it's been
- 16 collected.
- The problem is that there's basically
- what I'll call a moral lag in the way we treat
- ¹⁹ data.
- What I mean by that is that we as a
- society are often very slow to realize that a
- 22 particular use of data is harmful, especially

- when it involves data of racial and ethnic
- minorities, LGBT people, and others who have
- 3 historically lacked political power.
- In fact, the two most prominent examples
- of this moral lag involve the Department of
- Defense, or formerly the Department of War.
- During World War II, Japanese Americans
- 8 volunteered information about themselves and
- ⁹ their families in the census. They volunteered
- that information under a statutory promise from
- the federal government that that data would
- 12 remain confidential. This was a use restriction.
- What happened? As you know, in 1942,
- 14 Congress waived the confidentiality provisions
- and the Department of War used detailed census
- data to monitor and relocate Japanese Americans
- to internment camps.
- After World War II a similar story
- unfolded for gay and lesbian service members.
- They were prohibited from serving openly, and so
- 21 many of them turned to military chaplains,
- 22 psychologists, physicians.

- Yet routinely and even after don't ask,
- don't tell, the military would use that
- 3 confidentially collected data to out and
- 4 dishonorably discharge LGBT service members.
- Now today with the benefit of hindsight
- 6 we recognize that these events are
- discrimination, but at the time the picture was
- 8 less clear for a lot of people.
- And that took a long time to change.
- 10 The census only acknowledged the full extent of
- wartime sharing of census data in 2007, and
- 12 Congress only repealed the ban on openly serving
- gay and lesbian service members in 2011. That
- was three years ago.
- So let me be clear, my point is not to
- 16 cast aspersions on the Department of Defense,
- 17 rather my point is that all of us as a society
- 18 are consistently slow to recognize what's a
- 19 harmful use of data when it comes to vulnerable
- 20 communities. It often takes us decades to figure
- that out. Far too often today's discrimination
- was yesterday's national security measure.

- What this means for our data and what
- this means for privacy is that we cannot solely
- ³ rely on use restrictions.
- 4 What this means is that collection
- 5 matters and the that simplest and most powerful
- 6 way to protect privacy is to limit data
- ⁷ collection, particularly for the government.
- I urge you to continue to protect that
- ⁹ core right. Thank you.
- MR. DEMPSEY: Thank you very much. Our
- 11 next witness is Mike Hintze, who is Chief Privacy
- 12 Counsel at Microsoft, where he's been for sixteen
- and a half years at the epicenter of the
- evolution of technology and privacy.
- MR. HINTZE: Thank you. Thank you for
- the opportunity to speak with you today and
- participate in this important discussion.
- I come to this discussion from the
- 19 perspective advising on and managing privacy and
- related issues in the private sector.
- I've done that for nearly two decades,
- first as an associate here in a D.C. law firm,

- and as you mentioned for the last sixteen-plus
- years at Microsoft.
- At Microsoft we approach the issue of
- 4 privacy from a core belief that privacy is an
- ⁵ essential value, both to us and to our customers.
- ⁶ We have a strong commitment to privacy because we
- ⁷ recognize that customer trust is critical to the
- ⁸ adoption of online and cloud services.
- 9 Our customers, from individual consumers
- to large enterprises, will not use our products
- and services unless they trust them, unless they
- trust that their private data will remain
- ¹³ private.
- We seek to build that trust with our
- 15 customers by adhering to a robust set of policies
- 16 and standards. These policies and standards
- guide how we do business and how we design our
- 18 products and services in a way that protects
- 19 customer privacy.
- These standards are based on the Fair
- 21 Information Practices, which we agree remain
- relevant today, including transparency about the

- data and how we use it, minimization with regard
- to the data collected and how long it's retained,
- 3 choice about collection and use of data, strong
- 4 security to ensure that the data is protected,
- 5 and accountability to ensure that we are living
- 6 up to our commitments.
- 7 These standards are not just a rule that
- ⁸ we create it and hope that our employees follow.
- 9 Instead, we built them into the processes we use
- to operate our business.
- For example, they're built into the
- 12 tools that are used in our software development
- 13 life cycle, and there are checkpoints that
- 14 prevent a product or service from shipping
- without a privacy sign off.
- In sum, we've taken what's often
- 17 referred to as a privacy by design approach to
- 18 how we operate the company and how we develop and
- 19 run our services.
- 20 And this approach is supported by a
- 21 mature privacy program that includes dedicated
- personnel with privacy expertise who sit in both

- centralized roles and are embedded throughout the
- business. The program includes incident
- management, response and escalation processes.
- Further, we've developed and deployed
- 5 comprehensive role-based training for engineers,
- sales and marketing personnel, as well as those
- ⁷ in HR, customer service and other roles that
- 8 touch and handle personal data. And our program
- 9 includes executive level accountability for
- 10 privacy compliance.
- But that investment in privacy and the
- trust we've worked to build is undermined if
- those customers believe the government can freely
- 14 access that data.
- Concern about government access to data
- collected by the private sector can foster a lack
- of trust in those private sector services.
- And when those concerns are focused on
- the access to data by the U.S. government, that
- lack of trust becomes focused on U.S. companies.
- That's why we've been vocal for the need
- for surveillance reform in the United States.

- 1 There have been positive steps in this regard in
- the last year but there's more that needs to be
- 3 done.
- We've laid out several things the U.S.
- 5 government should do to help restore the trust
- that's been damaged by last year's revelations.
- First, bulk data collection programs
- 8 should end. We have been clear that we have not
- 9 received any bulk orders, any orders for bulk
- data collection, but we strongly feel that
- surveillance should be focused on specific
- 12 targets rather than bulk collection of data
- 13 related to ordinary people's activities and
- 14 communications.
- The recommendations of this Board on the
- Section 215 program are encouraging, as are the
- comments of the President, and we urge the
- administration to end the existing program, and
- we urge Congress to enact prohibitions on any
- such orders in the future.
- Second, we should do more to increase
- transparency. Transparency is a key element to

- any program for protecting privacy. It
- ² facilitates accountability and enables public
- debate around policies and programs.
- 4 Here too we've seen positive
- ⁵ developments. In particular, the government has
- 6 declassified more information about its
- ⁷ surveillance programs and the workings of the
- 8 FISA court.
- Additionally, we and other companies
- filed lawsuits last year against the U.S.
- 11 government arguing that we have a legal and a
- constitutional right to disclose more detailed
- information about the demands we've received
- under U.S. national security laws.
- And earlier this year we came to an
- agreement with the government enabling us to
- 17 publish some aggregated data about the FISA
- orders and the national security letters we've
- 19 received.
- It was a good step that helped foster
- better understanding of the type and volume of
- such orders that service providers received, but

- we believe there can and should be more detailed
- ² reporting permitted.
- Third, we support reforms of how the
- ⁴ FISA court operates. In order to foster a
- ⁵ greater confidence in surveillance programs and
- 6 government access to data that are appropriately
- balanced against privacy and other individual
- 8 rights, surveillance activities must be subject
- ⁹ to judicial oversight.
- We need a continued increase in the
- transparency of the FISA court's proceedings and
- 12 rulings, but effective judicial review requires a
- 13 true adversarial process where more than one side
- is heard. We urge Congress to act on FISA
- 15 reform.
- Fourth, government should provide
- 17 assurances that it will not attempt to hack into
- data centers and cables.
- 19 In the year since the Washington Post
- 20 reported an alleged hacking by the NSA of cables
- running between data centers of some of our
- competitors, there's not yet been any public

- commitment by the government that it will not
- seek to obtain data by hacking into Internet
- 3 companies.
- We believe the Constitution requires
- that the government seek information from
- 6 American companies within the rule of law and
- ⁷ through authorized government access, and we've
- 8 taken steps to prevent such attempts by
- 9 increasing and strengthening our use of
- encryption across our networks and services.
- Nevertheless, we and others in industry
- will continue to press for clear government
- assurances.
- Fifth, although recent government
- revelations have focused mainly on the U.S.
- government and many of the subsequent debates
- have focused on the privacy rights of U.S.
- persons, we must recognize that this is a global
- 19 issue.
- 20 As we seek to sell our products and
- services to customers around the world,
- discussions that focus exclusively on the rights

- of U.S. persons are not enough. Many people
- ² around the world do view privacy as a fundamental
- human right, and they have a very real concern
- 4 about whether and how governments can access that
- 5 data.
- In that regard, we appreciate the steps
- ⁷ that President Obama announced in January which
- 8 acknowledged the need to address protections
- 9 about non-U.S. citizens.
- Along those lines in the law enforcement
- 11 context, we've challenged a federal warrant in
- the U.S. courts seeking customer email for
- content that's held in our data center in
- 14 Ireland.
- Further, we've called for governments to
- come together to create a new international legal
- 17 framework that allows for new streamlined
- 18 processes for cross border data access that can
- supplement existing rules.
- None of this should be taken to suggest
- that we don't value and appreciate the absolutely
- 22 critical work that our law enforcement security

- agencies do every day to keep us all safe.
- In fact, we work closely with the U.S.
- and other governments to help fight cyber crime
- 4 and other threats. We want to ensure that those
- 5 agencies have the tools and information that they
- 6 need to protect us from terrorism and other
- ⁷ threats to our safety and security, but there
- ⁸ needs to be a balance between safety and the
- 9 personal freedoms that people around the world,
- 10 especially law-abiding citizens and institutions
- 11 enjoy.
- 12 This balance is rarely an easy one. As
- 13 Chief Justice Roberts recognized recently in the
- 14 case of Riley v. California, privacy comes at a
- 15 cost. But the court's unanimous decision makes
- 16 clear privacy is an inherent and enduring value
- that must be protected.
- While there's not always a perfect
- analogy between protecting privacy in the private
- sector, law enforcement, and national security
- 21 context, we also, we in the private sector
- regularly deal with questions of striking the

- 1 right balance between privacy and other needs.
- In each of these contexts as technology
- ³ evolves we need to continually reevaluate that
- 4 balance and many of the principles that have
- 5 proved useful in striking and retaining that
- balance, the Fair Information Principles,
- ⁷ continue to be relevant today.
- MR. DEMPSEY: Mike, could you wrap up?
- 9 MR. HINTZE: Thank you. I'll end my
- 10 comments there.
- MR. DEMPSEY: Okay, super, thanks.
- We'll come back to some of those issues with the
- 13 questions.
- Our final member of this panel is Hadi
- Nahari. He is Chief Security Architect at
- NVIDIA, a company that designs and builds high
- 17 performance computer systems. Hadi is a
- 18 cryptographer and computer scientist. Welcome,
- 19 please proceed.
- MR. NAHARI: Thanks for the opportunity
- to testify today. I appreciate it.
- I'm here as a technologist and not as a

- 1 lawyer. In Silicon Valley we say the "I'm not a
- 2 lawyer" rule applies.
- Our concern is about building systems
- 4 that are buildable and creating rules that are
- ⁵ enforceable, so I wish to provide some technology
- background to the panel and to the conversation.
- From our perspective security is to a
- 8 system what harmony is to music. Providing
- 9 security as a foundation of establishing rules of
- 10 privacy is our model.
- We build systems that are enabled and
- 12 are able to enforce rules, and that is the
- 13 context of security as we see it.
- Security is one of the intersections
- between technology and civil liberty, and we deal
- with issues such as trust and active adversary in
- 17 a system. This is how we build and design our
- systems.
- Our world used to be simpler, and
- sometimes I provide samples of that simpler
- world. You all remember this as a mobile phone.
- This is from the time that the phones were

- actually doing just that, they were a phone.
- 2 And some of these devices were
- 3 statements of class. You all remember this,
- 4 right? This was a phone. This was a mobile
- ⁵ phone. I worked in this company.
- One of my favorites in the collection,
- ⁷ text, this used to send and receive even text
- 8 messages.
- Oh, yeah, CLIE, this was your personal
- and digital assistant.
- I have some others. Oh, yeah, Palm,
- they used to be a company that existed, this was
- one of the darlings of the valley.
- So these, of course this was also a very
- important device that everyone carried.
- This is from the time that the world was
- very simple and we built systems that did very
- 18 basic things.
- And it was per Thomas Friedman, and I
- quote here, "When I sat down to write, The World
- is Flat, Facebook didn't exist, Twitter was still
- 22 a sound, the cloud was still in the sky, 4G was a

- parking place, LinkedIn was a prison,
- ² applications were what you sent to college, and
- 3 Skype was a typo."
- So June 29th, 2007, iPhone was
- introduced, the world changed. The world for us
- technologists changed, probably for everybody
- ⁷ else in the room, non-technologist and
- 8 technologists alike also changed. And we are
- 9 dealing with devices that are not as simple as
- what we used to carry.
- So that's part of the problem from my
- 12 perspective. I'm interested in the ramifications
- of the changes in this technology as the subject
- that we are talking about. It's only seven and a
- 15 half years.
- 16 It's only seven and a half years ago.
- 17 So I don't believe there's any other event in
- 18 history that in this short amount of time has
- 19 ravaged and gone through everything and tried to
- change everything, such as the foundation of our
- society.
- In the old and pre-2007 world we said

- things like, you cannot enumerate all the attacks
- in cryptography is a known statement. And state
- 3 space combinatorial explosion, meaning you cannot
- 4 define a secure state of a system. It was
- ⁵ difficult back then during these devices. It has
- ⁶ just become worse.
- The guarantees, we do not know anything
- 8 about our future but a couple of things I could
- guarantee, a couple of things I could guarantee
- right here is that things will only get faster.
- We're going to build things that are faster.
- 12 They're going to become smaller, a lot smaller.
- 13 They're going to become cheaper, and these
- devices are going to become a lot more abundant.
- Some of them, we no longer care about
- building devices that are usable for a long
- 17 period of time. It's a lot more economic to
- build these devices that are basically throwaway
- devices. That's the concept that we are
- following.
- And they're becoming more connected.
- 22 Everything is becoming more connected. You have

- 1 heard things such as IOT, Internet of things, or
- ² as I call them, thingsternet.
- Everything is just becoming very
- ⁴ talkative. All of these devices are very chatty.
- 5 They talk a lot. So you guys all have phones,
- 6 smartphones in your pockets. From the time that
- ⁷ I started, which was about five minutes right
- 8 now, until now, each one of those devices,
- ⁹ without you even touching them, has transmitted,
- sent and received, about half a meg data, without
- 11 you even touching them.
- 12 This abundance of information that is
- 13 happening that is, without you interacting, is
- having a lot of ramifications on what we are
- doing.
- We heard a lot of things about data is
- only, you know, accumulating. It's not going
- away. We are generating more data than we can
- manage or fathom.
- A hundred hours of video, a hundred
- 21 hours of video is uploaded on YouTube, and
- YouTube is not the only recipient of the service,

- other companies also have these services, a
- hundred hours of video are uploaded to YouTube
- every single minute. Every single minute.
- So we are building systems to manage,
- and compartmentalize, and define, and create and
- 6 work with these data. And this data, as we have
- ⁷ heard in the two panels, are not going away.
- 8 They are not disappearing.
- In the new world, maintaining security
- is even harder. So as a citizen, I'm very
- 11 carefully following what is happening by this
- 12 esteemed Board as to what is the ramification of
- 13 the decisions that we are making and whether
- that's enforceable, whether we can build systems
- that are enforcing these rules.
- Because right now being a security
- 17 professional and creating doable and enforceable
- 18 security is as unpopular as being an atheist in
- 19 Jerusalem. No one likes you. So I'm hoping that
- we can come up with a system that is also
- ²¹ buildable.
- 22 And lastly, I close my remarks and I'm

- 1 looking forward to the questions.
- One more thing that I could guarantee is
- the attacks are going to increase only, and
- 4 they're going to become simpler and easier to
- 5 mount.
- By one measure the number of attacks in
- ⁷ 2013 were three trillion, only affecting private
- information, on average 27.3 dollars per attack,
- 9 about a hundred billion dollars, the cost of
- these attacks. This data is 2013. None of the
- 11 Target, Home Depot, LinkedIn, none of that
- information, none of those attacks are included
- here.
- So with that, I close my remarks and I
- 15 look forward to answering questions. Thank you.
- MR. DEMPSEY: Thank you. We'll now go
- through a round of questioning, and Board members
- as well will be subject to the time limits here.
- 19 I think I have 20 minutes and then each Board
- member will have five minutes, and then still the
- 21 possibility of questions from members of the
- ²² audience.

- I wanted to build my first question off
- of the point that I think Hadi was making at the
- 9 end, which is that there seems to be this
- 4 inexorable trend towards more sophisticated
- ⁵ devices collecting, generating, sharing, emitting
- 6 autonomously, automatically disclosing more and
- ⁷ more information.
- And I think I'll go to Professor Anton
- ⁹ first and then maybe come back to Hadi with this,
- but looking at that phenomenon and the seeming
- inexorability of it, the seeming inevitability of
- it, first on the technology design side and then
- on the policy side, on the technology design side
- 14 what do you see as any potential at all for
- limiting that growth, controlling the flow of
- that information?
- You talked to some extent about the
- 18 possibility of technology protecting privacy.
- 19 How does that square with this tremendous ongoing
- 20 growth of information?
- MS. ANTON: Thank you. So you know, as
- was mentioned in the earlier panel, systems are

- getting more and more complex, which makes
- 2 compliance more and more difficult as well.
- I really hope that we don't limit growth
- 4 and limit the ingenuity of new technologies that
- 5 might have really great applications in the
- future and solve wonderful, really important
- ⁷ problems.
- By the same token, there is a lot of
- 9 work that's been done, especially with work
- that's being done at Georgia Tech, in fact, on
- 11 how do we design the Internet of things or the
- 12 Internet of devices, such that we are taking
- 13 privacy and security into consideration, give all
- 14 of the outputs, all of the possible inputs.
- And engineers just simply need better
- tools and heuristics for how to do that. And,
- you know, it's privacy by design, it's thinking
- about these things early on and not thinking
- 19 about it after the fact.
- 20 And in terms of controlling information,
- I think what we want is to secure the flow of
- information, but not limit the flow of

- ¹ information.
- 2 And these are all things that
- ³ researchers are actively working on in
- 4 universities and at research labs in industry as
- 5 well.
- MR. DEMPSEY: You know, I've written
- myself about the potential for privacy enhancing
- 8 technology, value of privacy by design. But at
- ⁹ the same time, I mean at some level I just don't
- see it happening.
- MS. ANTON: So --
- MR. DEMPSEY: Or let me put it this way,
- while I see it happening, and I think Mike
- Hintze's point that Microsoft has incorporated
- privacy by design as a corporate concept, but
- there are these other hugely dominant trends that
- 17 almost seem to be overwhelming.
- MS. ANTON: So within the context of
- 19 counterterrorism I think that there's a lot of
- 20 policies and a lot of laws that are in place.
- When I mentioned earlier that I'd like
- to see more technologists in the room, it's not

- just to kind of study it after the fact, but
- ² actually to be involved in forming the policy.
- Because a lot of times the policy and the law are
- written in such a way that we can't implement it.
- 5 And so what I'd like to see is more
- 6 technologists involved in the discussion up front
- 7 really informing the decisions about laws that
- are going to be passed, about the policies that
- ⁹ we're going to adopt, because we could write them
- in a way that makes it a lot easier to comply
- ¹¹ with the law.
- MR. DEMPSEY: Do you have an example in
- 13 mind?
- MS. ANTON: Excuse me?
- MR. DEMPSEY: Do you have an example in
- $16 \quad mind?$
- MS. ANTON: So I work a lot in HIPAA,
- 18 for instance. We have the new change with
- meaningful use. I had one Ph.D. student who was
- really working actively on how do we predict what
- the change is actually going to be? Because when
- they finally make that decision we're going to

- 1 have very little time to implement that change in
- 2 systems to be able to be able to make sure that
- 3 we're compliant with it.
- 4 And had we had more technologists
- involved in that process, we'd be able to more
- quickly adapt our systems and we'd have a better
- 7 community of practice, if you will, about how to
- 8 establish those laws and how to then instrument
- 9 systems to make sure that only the right people
- are having access to the right information at the
- 11 right time and in compliance with law.
- MR. DEMPSEY: Just to round that out,
- 13 certainly you would agree that we need both
- better, clearer laws, as well as more mindful
- technology design?
- MS. ANTON: Absolutely.
- MR. DEMPSEY: That it's not that one or
- the other will solve this problem.
- MS. ANTON: Absolutely, we need both,
- ²⁰ right.
- MR. DEMPSEY: I want to go to Alvaro
- Bedoya. There was one point in your written

- testimony that you didn't mention and I want you
- ² to talk about it now. I think it's very
- 3 important.
- 4 A lot of our constitutional law of
- ⁵ privacy is based upon the concept of reasonable
- 6 expectation of privacy. And there's a lot of
- worry and a lot of, I think, legitimate concern
- 8 that with these changes in technology that our
- 9 expectations of privacy diminish.
- You talked about the fact that, in fact,
- with changes in technology our expectations of
- 12 privacy may actually be growing. Could you
- 13 explain that?
- MR. BEDOYA: Yeah, that's exactly right.
- 15 And the point here is that the Katz test cuts
- both ways. You know, usually when the court
- talks about Katz in society, they say, well,
- everyone's becoming inured to this idea. They're
- 19 surrendering to the ubiquitous collection of
- 20 their data.
- But I actually think people are,
- technology is helping people learn about what

- they think privacy is.
- 2 And the best example of this I think is
- 3 location technology and facial recognition
- 4 technology.
- 5 Previously people had no occasion to
- 6 develop an opinion on whether or not they
- ⁷ expected, you know, the sum total of their
- 8 movements to be developed, to be compiled in a
- 9 profile, but suddenly it's becoming radically
- 10 cheaper to conduct that surveillance.
- And so I think that in the same ways
- that you only realize what you had when you start
- losing it, for the first time a reasonable
- 14 expectation of privacy in public is crystallizing
- in people's minds.
- And so I would say that ubiquitous
- surveillance is making people say, hey, you know
- 18 what, maybe when I go to the grocery store, or I
- drive down the street, or I go to work I expect
- 20 my colleagues at work to see me, you know, the
- 21 people I know at the store to see me, my
- neighbors to see me, but I really don't expect

- anyone to know that I'm at all those places at
- ² all times no matter where I go.
- And so I do think that technology can
- 4 expand our expectation of privacy.
- MR. DEMPSEY: And Mike Hintze, certainly
- over the past fifteen or sixteen years that
- you've been at Microsoft, do you think it's fair
- 8 to say that your customers have become less
- 9 interested and less concerned about privacy or
- expect more of Microsoft and other companies when
- 11 it comes to privacy?
- MR. HINTZE: I think they expect more.
- 13 I think, you know, I agree that expectations of
- 14 privacy in some ways have increased. They've
- 15 certainly changed.
- 16 As technology evolves people learn about
- it, they adapt. There's certainly data sharing
- going on that people wouldn't have contemplated
- or accepted a number of years ago, but that
- doesn't mean people don't care about privacy
- anymore.
- It's very clear to us that our customers

- care about privacy now more than ever. And you
- see that in the amount of resources and attention
- and focus that we've put on privacy.
- It really is one of the top legal issues
- we're dealing with. It's one of the top customer
- issues we're dealing with. We hear every day
- ⁷ from customers who have questions about how their
- data is being treated, how it's being protected,
- 9 how it's used. People's expectations of privacy
- are not fading away.
- MR. DEMPSEY: And by the way, just to
- 12 put a sort of nail in the coffin here, I think
- the government argues, and there's obviously
- Supreme Court precedent to support it, that a
- person surrenders his privacy rights when he
- discloses information to a third party such as
- 17 Microsoft in the course of using the Microsoft
- 18 products or services.
- But it seems to me from what you're
- saying that Microsoft does not believe that its
- 21 customers have surrendered their privacy rights
- when they use the Microsoft product or service,

- and thereby Microsoft has acquired information,
- ² Microsoft does not believe that that information
- ³ has zero privacy interests.
- MR. HINTZE: Absolutely not. On the
- ⁵ contrary. I mean to the extent that the third-
- party doctrine ever made any sense, it doesn't
- ⁷ make any sense today.
- I mean people increasingly are putting
- ⁹ all of the information that they used to keep in
- their homes, in their file cabinets, online in
- 11 cloud services.
- And as recent court decisions have
- recognized, particularly in Riley, it's even more
- data. There's more data in the cloud. There's
- more data being created that reveal the most
- private and intimate details of people's lives
- that's in cloud services in the hands of third
- parties, more so than was ever in people's homes.
- And the expectations around privacy
- around that data are quite profound.
- MR. DEMPSEY: And that's true, in your
- view, both of content, so to speak, and

- non-content, or metadata, or transactional data.
- ² There's sensitivity there in both categories.
- MR. HINTZE: Absolutely. You know I
- 4 don't like the term metadata because it
- ⁵ encompasses too much. I think we should talk
- 6 about what we're talking about.
- And you know, there's a broad range of
- data that's collected, or even created, or
- 9 inferred through the use of online service. And
- some of it's fairly benign.
- You know, we call things metadata, put
- the metadata label on things like the amount of
- 13 storage you're using in your online storage thing
- or the average file size, but even that has
- privacy implications. And we embrace the ideas
- of transparency, and consent, and all of the
- 17 FIPPs around that kind of data, too.
- But as you go up the scale with maybe
- 19 content being the end as sort of the most
- 20 private, the stuff that people have the highest
- 21 expectation of privacy around.
- But other things about who you're

- communicating with are right up there, right up
- ² against content in terms of what that can reveal
- about people's relationships, associations,
- thoughts, beliefs, etcetera. And there's very
- 5 important privacy implications around that data
- 6 as well.
- 7 MR. DEMPSEY: You mentioned the
- 8 trans-border issues and the fact that people
- ⁹ around the world recognize privacy as an
- interest, and in many cases as a human right.
- Just where do we stand and what are you
- aware of, or what do you know about, is there any
- 13 progress being made multilaterally, or
- bilaterally in terms of developing standards for
- trans-border surveillance and trans-border
- government access, anything in the works there
- that we should be aware of?
- MR. HINTZE: Not that I'm aware of
- 19 specifically. You know, there's certainly more
- discussions happening in recent years than there
- has been in the past around a number of
- 22 constituents and interested parties on privacy

- ¹ around the globe.
- Jim and I were recently at an
- international data protection conference where
- these issues were loudly and vigorously discussed
- 5 and debated.
- And so that dialogue is happening, but
- ⁷ in terms of actual progress towards making
- 8 headway in terms of developing an international
- ⁹ framework for this stuff, there's certainly a lot
- more work to be done.
- MR. DEMPSEY: May I just ask you and
- others, as well as members of the audience,
- additional panelists, if and when you do become
- aware of things that are making progress, please
- 15 let us know. Obviously we're remaining
- interested in that trans-border question.
- For Hadi Nahari, you know we've talked
- about privacy by design. In your experience do
- 19 technologists give adequate consideration to
- 20 privacy as they design products? And what more
- 21 could be done to encourage or promote privacy by
- design?

- MR. NAHARI: In technology we build
- things that are reasonably well-defined. So I
- ³ recognize in the previous panel there was a
- 4 discussion that you don't necessarily need to
- ⁵ define privacy to be able to enforce it.
- On the technology side, if we are able
- ⁷ to build a model that represents a need, then we
- ⁸ are very good at building it.
- I think part of the reason that mapping
- a very human, a very societal concept such as
- privacy into the devices that we build, the
- 12 services that we build and we use, sometimes it's
- simpler, sometimes it's not.
- To answer your question, I see a great
- deal of attention, a great deal of interest in
- the notion of privacy, privacy by design, secure
- by design, trustworthy by design.
- And especially in the field that we are
- dealing with, our model, in security of the
- device when we release it and goes to the field
- is a mutually distrusting system. So you don't
- 22 really know.

- Let me take a step back. It's one thing
- to build a server that resides in someone's data
- 3 center where you have full control over the
- 4 actual device and you have to control the flow of
- information, the software that is there and how
- 6 it's used.
- 7 It's another thing to build a device and
- 8 leave it in the hands of the users and guessing
- ⁹ what they want to do.
- And then it's one thing to have a notion
- of privacy, as we do, and build a system based on
- 12 that.
- 13 It's another thing when you take a look
- 14 at this, should I call it a generation gap as
- to -- there's this company called Snapchat and
- they had promised that whatever picture you take,
- it's going to disappear.
- Anyone who has worked in technology
- 19 knows things like this are not possible, you
- could simply just take a picture of that device.
- 21 But we call it job security.
- Then when they realize that this is not

- 1 really possible they announced it, and they are
- under the oversight of the government for about
- ³ 20, I think, years to make sure that they do
- 4 things right. And they are paying attention. I
- 5 know they are paying a lot of attention to make
- 6 sure they get things rights.
- But then you take a look at the users.
- 8 I think the stat was released last week or the
- ⁹ week before that they asked college students, 50
- percent, more than 50 percent of college students
- said, yeah, we still will use Snapchat. They are
- 12 aware. They understand.
- I don't know how to reconcile that.
- There is a new generation that has, I don't know
- whether it's a more or less, but certainly a
- different expectation and definition of privacy.
- 17 And there is a vagueness of what does that mean
- in terms of a system that could be built.
- Once those are, you know, in a
- reasonable state, we are really good at building
- 21 systems that satisfy those rules.
- Hence my opening remarks as to our model

- in the industry and in technology is we
- understand the rules, we are very good at, you
- know, creating those rules and building systems,
- devices, services and everything that enforce
- those rules, but it has to be buildable and it
- 6 has to be enforceable. The attention is
- ⁷ certainly there.
- MR. DEMPSEY: But the first premise is
- the rules have to be clear and if they're not
- clear, then you don't know what to build.
- MR. NAHARI: Semi-clear will do. We
- used to live in a world before 2007 that
- everything had to be really, really well-defined.
- 14 It no longer exists.
- We have a new generation of hackers that
- do not abide by the rules, therefore we have to
- create systems that are almost right. We are
- seeing it in the program languages, we are seeing
- it in the design of the system, we are seeing it
- in self-correcting systems. Sometimes somewhat
- ²¹ accurate will do.
- MR. DEMPSEY: Do you want to respond to

- 1 that?
- MS. ANTON: Sure. So this reminds me a
- 3 little bit about what I was talking about
- ⁴ practical encryption and anonymization. And so I
- 5 think there are times in certain applications
- 6 where that kind of risk is fine and there are
- other instances where it's not fine.
- And then that's where guidance from
- 9 PCLOB can be very helpful in terms of trying to
- figure out what are the risk profiles and when is
- it that we can have pretty good rules and when do
- we have to have very, very tight, accurate,
- 13 hundred percent certainty kind of rules.
- MR. DEMPSEY: Okay, thank you. At this
- point other members of the Board will pose some
- questions under the five minute rule. And we'll
- go in sort of reverse order down the line here
- 18 starting with Rachel Brand.
- MS. BRAND: Thank you, Jim, and thanks
- to all of you for being here.
- That's actually a really good segue
- because the first question I was planning to ask

- was, Dr. Anton, I was interested in what you were
- saying about not letting the perfect be the enemy
- of the good in terms of de-identification. In
- 4 the domestic violence context you want it to be
- ⁵ perfect perhaps, and in other contexts good
- 6 enough will do.
- 7 Can you explain what you mean by that?
- 8 What's an example of a de-identification method
- ⁹ that might be good enough but perhaps not
- 10 perfect?
- I'm not a technologist, as you know, so
- if you can help me out, that'd be great.
- MS. ANTON: All right. So there are
- 14 certain cases of studies that have been done, for
- instance, when the Netflix put out their data
- online and then researchers went and looked at
- the Internet Movie Database to try to see whether
- they could re-identify people. They had
- resources, it was readily available information.
- 20 In this context I don't think anyone was
- 21 personally hurt by it.
- But there might be cases where that kind

- of identification could be extremely damaging.
- And so the more, we talked earlier about
- 3 aggregation of databases and how the ability to
- 4 link different kinds of information across
- ⁵ different kinds of databases could actually be
- 6 detrimental. It can also help us find the bad
- ⁷ guy though. And so that's the tension, right?
- 8 So when is it okay and when is it not
- 9 okay? And are there instances, for instance, for
- Netflix or something that's available online
- that's just not, you know, where you went to
- school or something that's not very important.
- 13 It may not be really necessary to worry about
- where you had dinner, for instance.
- But in a context of a group that is
- actively trying to announce a terrorist attack,
- then that's really important.
- MS. BRAND: So I guess that makes sense
- in terms of when it's important and when it's not
- important, but how do you do it? I mean like,
- for example, how do you do the perfect in the
- ²² domestic violence context?

- MS. ANTON: I think that's very
- difficult. I think we have technology that's
- 3 pretty good but not perfect. And so the idea is
- 4 do you keep the data unencrypted and then easily
- 5 accessible, because it's not very important, or
- 6 do you actually encrypt it and then use
- ⁷ reasonable, practicable anonymization on top of
- 8 that?
- 9 So it just depends. And I think this is
- one of those cases where technologists would
- welcome guidance in helping us to figure out what
- 12 are the risk profiles, because technologists
- don't have access to sometimes what the risks are
- within a counterterrorism context.
- MS. BRAND: For Mr. Bedoya, you said
- something along the lines of, in the national
- security context some of the FIPPs must apply
- even if they all can't. Can you elaborate a
- 19 little bit more?
- MR. BEDOYA: Yeah, sure. So the first
- is a historical point, which is that when the HEW
- report was issued, I was just reading, it's like

- 1 pages 74, 75, the committee actually says, okay,
- we just set out these standards, clearly all of
- them can't apply to all intelligence records, but
- 4 some of them must apply because the risk is too
- 5 high if we don't have some protections.
- So to put that more concretely,
- obviously the difficult ones are individual
- 8 participation and transparency. And I think
- ⁹ there are ways to address these, at least on an
- aggregate level that would be really powerful.
- So, you know, I think in the 702 context
- 12 the Board has -- and to take a step back, I think
- 13 it is shocking that one and a half years after
- the Snowden disclosures the American public
- doesn't have even a rough sense of how many of
- them have had their information collected.
- Take the telephone records program,
- people think it's everyone, but then you have
- 19 news reports saying actually only 30 percent of
- calls are actually recorded.
- And so in the 702 context the Board has
- recommended various measures to identify the

- scope. In all my time in the Senate I never saw
- anything that would lead me to believe that it
- would actually be impossible for the NSA to
- 4 produce an estimate based on statistical sampling
- of the number of U.S. persons collected in 702
- 6 data.
- In the 12333 context there's a number of
- 8 things you could do to quantify scope. One of
- ⁹ them could just be releasing the number of
- queries done on USP data and 12333 data.
- So I think there are ways to address
- these principles at the aggregate level, if not
- ¹³ at the individual level.
- MS. BRAND: Okay. Anybody else have a
- 15 thought?
- MS. ANTON: I have a thought on that in
- terms of transparency. This is another way in
- which, for instance, FISC technologists could be
- 19 helpful because when you have -- if Hadi whispers
- in Mike's ear, I spoke with Jim Dempsey about the
- panel, by the time that gets to Jim it's going to
- be, I spoke with Jim about wearing flannel. It's

- going to be something completely different.
- So when you get lawyers talking together
- from the NSA and the FISC about technology, and
- 4 you don't have a technologist there to ask
- ⁵ questions or make suggestions about, well, we
- 6 could actually, have you thought about including
- ⁷ this kind of metric, or collecting this kind of
- 8 data, or instrumenting the software in certain
- ⁹ ways, we could actually improve the ability to
- have more transparency and more oversight in
- technology with those discussions, bringing
- everyone in the room.
- MS. BRAND: Thank you.
- MR. DEMPSEY: Chairman Medine.
- MR. MEDINE: I'm going to try to get a
- question in for each panelist, so I'd appreciate
- brief responses.
- For Annie, you said something that
- surprised me a little bit, which is that
- encryption is good for counterterrorism.
- 21 And I guess I would like to understand
- more. I understand having or mandating a

- backdoor weakens protections, but why? It would
- seem as though terrorists can now hide their
- 3 communications, which seems to be detrimental to
- 4 counterterrorism.
- MS. ANTON: I think it's a better world
- 6 when everyone can hide their information. And so
- ⁷ there was a case in Greece where there was a
- 9 phone and someone was able to actually start,
- because of the backdoor and the known exploits,
- they were able to actually listen to the
- conversations, basically do a wiretap on the
- prime minister. That's what happens when you
- don't have encryption and you don't have security
- 14 by default.
- And so to think that the terrorists
- aren't going to do the same thing, I think is
- 17 naive.
- MR. MEDINE: Alvaro, you talked about
- the expectation of privacy, and if I heard you
- correctly, but tell me if I'm wrong, is that
- you're in a sense suggesting that we talk about
- not what people expect their privacy to be,

- because I can put up a sign saying I'm conducting
- video surveillance and I can destroy that, but
- 3 their expectations of what privacy should be, a
- 4 more normative standard.
- MR. BEDOYA: So I'm actually not saying
- that. So that's a separate wonderful, powerful
- ⁷ argument.
- What I'm saying is that technology is
- 9 making us realize that we do expect privacy in
- scenarios that didn't exist ten or 15 years ago.
- 11 So I think technology can expand our notion of
- 12 privacy.
- But I also think that the Fourth
- 14 Amendment doesn't just protect me and you, it
- protects us as a society and it sets a base for a
- 16 relationship between a government and its
- 17 citizens that also needs to be protected.
- MR. MEDINE: Okay. And I guess this is
- 19 for Mike, the Fourth Amendment, which is you
- talked about the balance between government
- 21 requests and your customers' privacy. Do you
- think the government should have a warrant every

- time it accesses your customers' records,
- particularly if they're American customers?
- MR. HINTZE: Yeah, I mean certainly in
- 4 the law enforcement context we've advocated for a
- ⁵ reform of that that would in effect require a
- 6 warrant for access to any content, regardless of
- ⁷ the age, to precise location information, other
- 8 sensitive data.
- 9 You know, I'm not sure we would go so
- far as to say that a warrant is required in every
- single case for every single data type, but we
- certainly need to update the rules so that there
- is appropriate judicial review of surveillance
- 14 programs and specific requests that we get for
- 15 data.
- MR. MEDINE: So in terms of the third-
- party doctrine, would you then essentially not
- have it be an absolute exception to the Fourth
- 19 Amendment, but essentially where would you go
- with it to provide some protection, but not
- 21 necessarily a full warrant protection?
- MR. HINTZE: Yeah, I mean the laws that

- we deal with in the law enforcement context
- provide a sliding scale, in effect.
- I mean 2703(d) orders provide some
- 4 reasonable oversight and protection, something
- below warrant and probable cause, and we've taken
- the position that that's appropriate for some
- ⁷ types of subscriber data, etcetera.
- MR. MEDINE: Thanks. And, Hadi, you
- ⁹ talked about, and I want to put this in the
- 10 context of how much information should be
- 11 collected, and you talked about enforceable rules
- for collection, but you also said that collection
- is going to be faster, cheaper, and we're going
- to be all more connected, and that attacks will
- increase, and that even compliance with rules may
- be more difficult.
- 17 Professor Felten talked about potential
- abuse of information and also the increased
- 19 possibilities of breach.
- How would you strike the balance between
- collection rules and essentially use rules?
- MR. NAHARI: That's a very difficult

- question, a very difficult one. I don't know if
- in the technology side of the house, I don't know
- if we really know where the balance is.
- We take a look at the attacks, we take a
- look at the system, we take a look at the
- 6 capabilities, we take a look at the mere fact
- ⁷ that all of these attacks, all of these exploits
- 8 are becoming so advanced that I used -- to give
- you one concrete example, I used to need to be
- physically around your things that you touched to
- be able to lift your fingerprint and then have
- 12 access to your phone and then use that
- 13 fingerprint to mount an attack and use your
- 14 biometry.
- With the resolution of the cameras that
- we have these days, sometimes with a very high
- 17 resolution camera, I just need to have your
- 18 picture that was taken somewhere in China to be
- 19 able to zoom and zoom and zoom and then lift your
- fingerprint and mount an attack.
- Now, how do you reflect things like this
- 22 as to should we build systems that whenever

- there's a fingerprint, it smudges it and we don't
- ² expose it? There are things like this that I
- ³ encompass all of those use cases as it should be
- ⁴ buildable.
- But what I'm trying to get across is
- 6 coming up with the rules that define those
- 7 capabilities or things that should be and
- 8 shouldn't be done is a very complex problem.
- 9 MR. MEDINE: Thank you.
- MS. COLLINS COOK: So thank you guys for
- another excellent panel.
- My first question, and this goes back to
- 13 what I had said on the previous panel, which is I
- view our job to be translating these ideas, these
- concepts, these concerns into practical
- 16 recommendations.
- So starting with you, Mr. Hintze, what
- have you found effective as a privacy officer to
- ensure your very large workforce, your
- 20 complicated workforce dealing with emerging
- issues takes privacy seriously, your rules are
- enforced, and that from beginning to end privacy

- is a part of your culture?
- Because we have a new NSA privacy
- officer, so this is free advice to the new
- ⁴ privacy officer over at NSA.
- MR. HINTZE: Well, thank you. You know,
- as I alluded to in my opening remarks, you know,
- one, there's no silver bullet. You need to take
- ⁸ a number of approaches.
- And we've taken a number of approaches
- to drive awareness and sensitivity around privacy
- throughout our workforce through a number of
- steps, some mandatory training that's required
- 13 for all employees that cover a range of ethical
- 14 and compliance issues, deeper role-based training
- that's specific to software engineers, that's
- specific to sales and marketing people, that's
- 17 specific to different roles that people play in
- the company that impact customer privacy.
- We have, as I mentioned, not just sort
- of told people what the rules are and then
- 21 crossed our fingers and hope they abide by them.
- We have put in checkpoints in the way

- that we have developed our internal systems, the
- way you develop a software and get it out the
- door that has to go through certain checkpoints
- 4 and reviews to ensure that privacy issues aren't
- 5 missed or overlooked.
- So there's a number of things we've done
- ⁷ along those lines to make sure that people are
- 8 aware and have the tools available to them to do
- ⁹ privacy right.
- But then there's also different checks
- along the way to ensure that mistakes don't get
- 12 made.
- And nothing's perfect of course, but we
- try to do a multifaceted approach, or a
- multi-layered approach to make sure that we catch
- those things.
- MS. COLLINS COOK: And so let me follow-
- up on this, and it's a somewhat specific example
- 19 but hypothetical.
- Have you found training to be more
- 21 effective or effective enough in the absence of
- pairing with mechanisms and processes?

- That was a horrible question, so I'm
- ² just going to start over again.
- So 702, that program has certain legal
- 4 requirements. In the privacy sector would you
- train to those legal requirements or would you
- 6 also have, for example, when an analyst is
- ⁷ sitting there attempting to target, or select, or
- 8 whatever they're going to do, also have at each
- 9 stage of the screen, or the process, or however
- they're doing it, rules reflected in the computer
- system that they're attempting to use?
- MR. HINTZE: We do both. To the extent
- that you can use technology to enforce policy
- that's always super effective because you get
- past or you reduce the potential for human error.
- But that's not always possible. You
- can't completely prevent mistakes, oversight, or
- intentional bad acts. And so you need to do more
- 19 than that.
- You have to have, you have to build the
- 21 awareness so that the inadvertent stuff is
- reduced. You have to build in the technology

- tools to try to prevent that from happening.
- 2 And then you need some level of checks
- ³ to make sure that everything went right. And if
- 4 it's, you know, somebody who's intentionally
- trying to circumvent a policy for whatever
- fereign reason, that there's some way to catch that
- ⁷ before it creates a negative impact.
- 8 MS. COLLINS COOK: So I think I have
- ⁹ time for one other quick question.
- In the Section 215 program one of the
- 11 features was, in fact, that not all of the call
- detail record went to the government. In fact,
- names are not provided originally to the
- 14 government, and subscriber information, simply
- numbers to numbers. Would that be an example of
- de-identification and anonymization?
- MR. ANTON: Sure.
- MS. COLLINS COOK: That was my only
- ¹⁹ question.
- MS. WALD: I have a couple of very sort
- of brief questions, which I think you can answer
- very quickly and that way I'll get them all in.

- I'll begin with Annie. You talked about
- how it would be good for us, and we already do
- ³ have technologists on the Board. I'll ask the
- 4 government when we have the government board
- bere, too, but let me ask you based upon your
- 6 knowledge here, does the government have
- ⁷ technologists who worry at all about privacy?
- I know they have technologists
- 9 obviously, but is this, as a result of your
- observations and study in the field, something
- that they consult with the technologists about,
- hey, we need this kind of information for
- 13 national security, but we'd like to get it or as
- much as we can, what's the balance? Does any of
- that kind of thing go on inside the government
- with technologists?
- MS. ANTON: Right. So having worked a
- 18 lot with the government I know that they consult
- 19 technologists greatly with security, with
- 20 privacy, with compliance issues, and how do we
- 21 engineer software that takes all of that into
- ²² consideration.

1 I think if we look at the past five 2 years or so, or six years or so that you'll see 3 that the NSA was really, really focused on compliance. I think the results of the reports 5 and the oversight has shown that they've done a 6 really good job with that. When there's been an issue, they've dealt with it. 8 I think someone mentioned the new CPO at 9 I think what we'll see different now is 10 that not only is the, are we complying with law, 11 going to be something that's factored into all of 12 the software that's developed and all of the 13 tools and the techniques and the procedures, but 14 also now well, just because it complies with law 15 should we really be doing it, and what's the 16 extra step we're going to take to really consider 17 privacy at the onset? 18 MS. WALD: So you sound reasonably 19 satisfied with the fact that they're taking it 20 seriously and doing the best they can? 21 MS. ANTON: I absolutely do. I wish, I 22 actually feel very comforted by the fact that the

- government has a ton of oversight and a ton of
- 2 laws to comply with.
- I personally am much more worried about
- 4 the large collection, amount of collection that's
- taking place in industry that people don't really
- 6 understand.
- MS. WALD: All right. So I can go on to
- 8 my next.
- 9 Mr. Bedoya, you talked about how
- important it was to limit collection to what was
- 11 necessary or purposeful, etcetera, but in light
- of so many of the experts on both panels have
- talked about almost like an almost inevitable
- momentum of collection, collection, collection,
- where would you look, what part of the government
- or where would you look for the mechanism to try
- and limit the collection, or get that kind of
- impediment or balance done?
- MR. BEDOYA: Certainly. So I think
- folks have been saying that it's inevitable that
- industry is going to collect all this data. I
- don't think folks have been saying that it's

- inevitable that government will collect it.
- And I, for one, don't actually think
- it's inevitable that industry will collect it.
- But taking that as a given, I think the
- ⁵ question is about reconstructing the firewall
- between government and industry with respect to
- ⁷ data collection.
- And so I'd be surprised if anyone on the
- 9 panel thinks, or on the previous panels thinks
- that it's inevitable the government will collect
- 11 all this data.
- One quick other point, Judge Wald, on
- your previous question, I should note that I
- believe that the congressional committees that
- conduct oversight on FISA and on foreign
- intelligence, certainly the Senate Judiciary
- 17 Committee lacks a technologist, and I think
- that's an issue that needs to be addressed.
- MS. WALD: I think we talked a little
- 20 bit about that in our first report on FISA
- 21 reform.
- Okay, Mr. Hintze, you talked earlier,

- you said one of your principles was there
- shouldn't be any bulk data collections.
- Now terminology is varied all over the
- ⁴ place, so it would help me if I knew what you
- meant by bulk collection there.
- And let me just tell you, one gathering
- of public health people and they talked about the
- great importance of public health data, you know,
- 9 especially for when epidemics come along or that
- 10 sort of stuff.
- So wouldn't some of that come under your
- ban against all bulk data collection?
- MR. HINTZE: I was talking specifically
- about government surveillance programs that come
- 15 to industry.
- MS. WALD: Okay, I just wanted to
- 17 clarify that because -- and what do you mean by,
- 18 give us an example of what you would call bulk
- 19 data. Because this has been a debate as to
- whether this program or that program falls under
- ²¹ bulk data.
- MR. HINTZE: Certainly. I had in mind

- the 215 program in particular where government
- ² goes to service providers.
- MS. WALD: Where it's not targeted?
- MR. HINTZE: Yes, it's not targeted,
- ⁵ correct.
- 6 MS. WALD: I think that's all I have
- ⁷ right now.
- MR. DEMPSEY: We may be able to go back
- ⁹ to Board members for additional questions. I
- would like to continue with this panel up until
- the top of the hour.
- We have one question from the audience
- which I will read, and we welcome others if
- others want to pose questions.
- In 2005, the National Academy of
- 16 Sciences studied whether pattern-based data
- mining can anticipate who was likely to be a
- 18 future terrorist. It concluded that this wasn't
- 19 feasible.
- And the question is, is pattern-based
- data mining in the terrorism context, is it
- feasible today and will it be feasible ten years

- from now? Would anybody like to address that?
- ² Hadi?
- MR. NAHARI: I don't know specifically
- 4 about terrorism. I'm mindful of what Ed
- mentioned is that we have limited data.
- But there is a program that has been
- 7 running in Los Angeles in LAPD. We may not
- 8 necessarily still be able to identify specific
- ⁹ criminals, but our predictive modeling systems
- have been at work. They're able to make a
- 11 reasonably good prediction about where the
- criminal activities are more likely.
- 13 It is not precisely the question that
- you're asking, but I can assure that it is just
- becoming better. I can assure that any service
- 16 provider that has the amount of data that we are
- generating, and it's becoming more and more and
- more generated, is just honing and fine tuning
- ¹⁹ and polishing their models.
- Whether it's going to be applicable to
- 21 antiterrorism methods, I don't know. I think all
- of these models are heavily data-driven. So one

- 1 would need a lot of data.
- But to the point that these models,
- 3 these predictive modeling are able to predict
- 4 things may relate indirectly to terrorism or
- ⁵ criminal activities, the systems are suggesting
- 6 that we are going that way.
- 7 MR. DEMPSEY: Other thoughts on that
- 9 question?
- There's a system in Chicago that the
- 10 Chicago Police Department has deployed, which
- both has been touted and criticized, but it does
- somewhat at the neighborhood or block level
- 13 predictive or predictions as to criminal activity
- as well as, I understand, individual level,
- identifying people who may be either victims of
- crimes or perpetrators of crimes. Again, both
- touted and highly criticized.
- Any thoughts or comments?
- MR. BEDOYA: One just quick one, which
- is the risk of creating a feedback loop. You
- 21 know, if you predict that there will be crime on
- corner X, you watch corner X like a hawk, you see

- every crime that occurs on corner X and you
- therefore draw an over-represented sample of
- 3 crimes at corner X, reinforcing your prior
- 4 conviction that you thought corner X was real
- 5 dangerous. So that's the main one from my
- ⁶ perspective.
- MS. ANTON: So this is certainly not
- necessarily my area of expertise, however
- ⁹ predictive is different from being able to
- 10 reconstruct after the fact. And so can we use
- these things to then, when something has
- happened, go back and find whether we missed
- 13 certain people that are still involved? Yes, I
- do believe that's the case.
- In terms of predictive, I think we have
- a ways to go. By the same token I get, every
- morning I get a crime ratings, a crime report for
- 18 all the crime in my area. And I can tell you, I
- can predict where there's going to be, on a
- weekly basis, crime in my neighborhood. So, you
- know, we're getting there.
- MR. DEMPSEY: Well, I mean on some level

- that's just Comstat all over again, the systems
- that have been available to police for decades.
- MS. ANTON: Sure.
- 4 MR. DEMPSEY: One question, and I'll go
- down the row again, and I'll pose the question
- and I think we can just go down the row with
- additional Board members if they have, the Board
- 8 members have additional questions.
- I had said in talking to each of the
- panelists that I didn't want this to be a panel
- 11 about going dark and the implications of
- encryption, but several of you have alluded to
- encryption and its significance here, and I would
- 14 ask any of you who would, to comment on the
- ¹⁵ following, which is, there is a growing trend
- towards more and more devices, cheaper and
- cheaper wearables, and the Internet of things,
- and more and more data collection occurring.
- There's also it seems a trend towards
- more encryption by default, whether it's at the
- device level or, as Mike Hintze was referring to
- in terms of the encryption of data flowing

- between data centers.
- So it seems to me like we have two
- things going on at once, which is not unusual.
- 4 Somebody referred to the modern era, the era of
- ⁵ the Internet of things, big data, ubiquitous data
- flows, as the golden age of surveillance.
- And it seems to me that both trends will
- 8 always be there. More and more information
- ⁹ available both to the private sector and possibly
- to the government, and increasing pervasiveness
- or at least increasing diffusion, if not
- comprehensive diffusion of encryption.
- Comments on that as a premise, first of
- all, the premise of my question, am I right?
- And then secondly, where does that leave
- the government, and would you agree with my
- 17 assumption that there will still be huge amounts
- of information available, both to the private
- sector for its purposes, as well as to the
- 20 government?
- I guess let's go right down the row.
- 22 Professor Anton?

- MS. ANTON: So I believe that there will
- still be a lot of data that's available to
- 3 government. When I say that I really support
- 4 encryption by default, I also really think that
- our country really, we were the code hackers, and
- it was really critical in World War II.
- And I think that instead of just kind of
- 8 taking the lazy approach and saying, oh, leave us
- ⁹ a backdoor, that we should just get better at
- cracking the code, because they're getting
- smarter and we need to get smarter, too.
- And so I leave it to the lawyers to
- decide what the legality of when you can actually
- apply that or break into a system.
- But being satisfied with just having a
- backdoor means that we're not advancing our state
- of the craft and our tradecraft here in this
- country and we're going to be left behind as a
- 19 result.
- MR. BEDOYA: I'll actually pass.
- MR. DEMPSEY: Yeah, my thoughts on this,
- two trends seem to be occurring simultaneously.

- MR. HINTZE: Yeah, I mean we're
- ² certainly seeing an expanded use of encryption,
- ³ encryption between customers and the service
- 4 provider, and encryption between data centers,
- ⁵ encryption on devices, etcetera.
- And that's being driven by customer
- ⁷ demand. I mean customers are concerned about the
- 8 security of their data. And they're not just
- 9 concerned about the security of their data
- vis-a-vis hackers and bad guys, they're
- increasingly concerned about the security of
- their data vis-a-vis governments.
- And so to the extent that there is that
- 14 concern out there that's driving customer demand
- 15 for these security features and companies will
- 16 continue to invest in that.
- Does that mean that there will be no
- 18 data available? I don't think so. I mean the
- 19 nature of many cloud services requires service
- 20 provider access to it.
- You can't run an effective email system
- without being able to filter the content for spam

- ¹ and malware.
- And so there will be a point in the
- 3 communication chain where data is available, and
- 4 that means that if it's available to a service
- provider, it's available to a government through
- 6 lawful demands. So I guess that's it.
- 7 MR. DEMPSEY: Hadi, any thoughts on
- 8 this, and then I'll yield.
- 9 MR. NAHARI: First off, I want to agree
- with Dr. Anton's point, we should just get
- better. We cannot ask industry, oh, don't
- encrypt, don't do anything. I would love to
- follow that when Chinese and Russians also follow
- that as well. So that's just not going to work.
- 15 I'm very respectful of the problems that
- the law enforcement agency has with the current
- state of affairs. We just have to get better.
- And it works, at the end it's going to
- work better for us as a nation. So that's number
- one, I fully agree.
- Some of the things, so going dark, I
- don't know if it's going dark. I know that we

- 1 are currently in a state that we are really able
- to think a certain way about the system design,
- 3 about the system security, about maintaining
- ⁴ privacy, that world has changed.
- 5 The world and the industry has changed
- 6 rapidly. The rest of us are catching up. So I
- ⁷ think it pays dividend if we figure out, take
- 8 some time, figure out what are the rules of this
- 9 new world where we don't necessarily need to rely
- only on encryption.
- I'm a big fan of encryption. I think
- it's one of the tools that security professionals
- 13 and everyone has, but there are others. The fact
- that some data is encrypted is not on its own
- necessarily the end of the world.
- I mean how many times, I know Michael
- mentioned that we are overusing this notion of
- metadata, but if you think about metadata as
- something about the data, it is meaningful when
- you see some encrypted data is being accessed a
- little bit more than the other. One could
- discern, one could learn things about it.

- Once we start learning how to deal with
- this system, then we could maintain encryption,
- then we could maintain stronger encryption. We
- 4 could also deal with the cases where we don't
- 5 have access to clear.
- I think our law enforcement, I think our
- ⁷ government, I think our legal system, I think us
- 8 as a society are in the process of learning how
- ⁹ to deal with this new world were things that we
- knew in the past no longer apply.
- Lastly, the new generation have figured
- it out. I think they're doing a lot better.
- 13 They're figuring out that you cannot expect
- 14 everything is going to be fully protected for
- you. They're figuring out ways to live in the
- world where they're posting a lot of things on
- Facebook that, I mean us probably won't do.
- They're trying to learn how to deal with
- a system that, you know, you may not have the
- 20 capabilities of asserting your privacy in the way
- that our generation did, but still have an
- expectation about their rights.

- MR. DEMPSEY: Does a particular Board
- 2 member have a question? Yes?
- MS. BRAND: Several of you have referred
- 4 to oversight in one way or another and I just
- want to ask a question about that.
- In my view, oversight is especially
- ⁷ important in the intelligence context because of
- 8 the necessary level of secrecy that attends.
- 9 It's important in all areas of government, but
- 10 especially here.
- But at the same time, when you start to
- layer on box checking exercises and paperwork
- 13 there is a point of diminishing returns and you
- sort of have oversight for its own sake that
- doesn't actually deter misconduct or ensure
- 16 compliance with the rules.
- Do any of you have thoughts on
- 18 principles for what's effective oversight, as
- opposed to just another box checking exercise?
- MR. BEDOYA: So I certainly have a few
- thoughts for the legislature. I think that
- there's been a lot of soul searching around how

- the executive needs a change in practices with
- ² respect to internal oversight. But I think
- there's some pretty serious problems at the
- 4 legislature.
- One of them is the technologist issue
- 6 that I mentioned. Another is clearances. I can
- ⁷ say with moderate to high confidence that most
- 8 United States senators lack a staffer with TSSI
- 9 clearance. I hope I'm wrong. I don't think I
- 10 am.
- And the fact is that all of the key
- 12 briefings for these senators are conducted at
- that level. And as a staffer, I know there's a
- 14 lot of staffers in the room, you don't send your
- boss into a meeting about soybeans without a TSSI
- staffer -- sorry, no, you don't need a TSSI
- staffer for that, but you don't send them into a
- meeting on an issue that seems very easy without
- a staffer. And a lot of these folks are going in
- on staff.
- Now thankfully folks on judiciary and
- intel have dedicated TSSI folks for the committee

- that they can rely on, but outside of those
- committees you're often flying, I don't want to
- 3 say flying blind, but you don't have the
- 4 resources you need to actually conduct that
- ⁵ serious oversight.
- 6 MR. MEDINE: I have two questions for
- ⁷ Professor Anton on de-identification.
- One is you commented earlier that phone
- 9 numbers without names associated with them would
- be de-identified information --
- MS. ANTON: It's actually not
- de-identified, because if it's my cell phone, I
- stand corrected on that.
- MR. MEDINE: Okay. Because obviously
- the availability of reverse directories makes
- ¹⁶ that --
- MS. ANTON: Absolutely, sorry.
- MR. MEDINE: Then I guess you also had
- 19 commented earlier that by analogy of having a
- lock on your door was a pretty good protection
- 21 against burglars but obviously not a perfect
- ²² protection.

- And I guess the question is, in the
- 2 context of a massive database burglars may not
- have the incentive or wherewithal to break into
- 4 everyone's home in a community, but with a
- 5 massive database with a brute force attack, you
- 6 might be able to get a very valuable return on
- ⁷ it.
- 8 So does that suggest that
- ⁹ de-identification needs to be essentially
- stronger or may not even be sufficient?
- I mean as you pointed out on the Netflix
- example, and Professors Paul Ohm and Latanya
- Sweeney have written articles about the ability
- to de-identify, is it a useful tool in some
- instances but not others?
- And even where it's useful, does it to
- have to be a pretty enhanced form of
- 18 de-identification?
- MS. ANTON: Well, I think it's better
- than nothing. You have to work harder at it to
- get access to it, right, and to really be able to
- understand it. But that's going to help us with

- the, you know, high school kid who's just trying
- ² to tinker around, right?
- But I think this is another example
- 4 where encryption is really, really important, and
- ⁵ very strong encryption. And so I think it's a
- 6 blend of both.
- 7 MR. MEDINE: Thank you.
- MS. COLLINS COOK: Just on the issue of
- 9 de-identification and anonymization, I had
- understood it as a concept that could apply in
- varying degrees. So at a period of time it has
- been de-linked from the identifying information
- and now they have to go to court in order to
- 14 re-associate it with the identifying information.
- So I don't think I was asking you to say
- that it had been permanently de-identified or
- ¹⁷ anonymized.
- This question is for Mr. Bedoya. To the
- extent that we're looking at evolving standards
- or evolving notions of expectations of privacy,
- 21 how do you quantify it?
- Is it because 51 percent of folks in a

- Washington Post poll said I care about this but
- I'm still using Facebook? Do you look at
- 3 conduct? Do you look at the fact that people
- 4 inside the beltway really care? People in ivy
- ⁵ leagues really care? I struggle with what is a
- 6 good way to identify emerging notions of
- ⁷ expectation of privacy.
- MR. BEDOYA: I'm not going to pretend to
- 9 know the right answer to that question. It's a
- really, really hard question.
- I certainly think that looking at
- 12 conduct is extremely valuable, and there's been a
- 13 lot of discussion about the third-party doctrine.
- 14 And the fact is it doesn't remotely represent
- what the American people think about privacy.
- You know, if your social network only
- 17 had the settings of public and only me, that was
- the only option, you know, people would say this
- 19 is ridiculous.
- And I do think it sounds strange to say
- it, but we do have something to learn from the
- best practices of these social networks, in that

- they very much see the world as a series of
- 2 segments and they respect the fact that sometimes
- you want to share something with segment A and
- 4 not segment B. And so I would say that's
- ⁵ certainly valuable.
- I don't have a good test about
- ⁷ identifying a reasonable expectation of privacy.
- 8 I'll just repeat myself in that I think we need
- ⁹ to see that as a standard that can expand and
- 10 contract.
- MS. ANTON: If I could quickly add,
- 12 after the Snowden leaks there's an anonymous
- search engine called DuckDuckGo and the number of
- people who started doing searches on that search
- engine increased, I think by over a hundred
- 16 percent. So there's one way that you can watch
- people's actions and conduct.
- MR. HINTZE: Just one very, very quick
- 19 add-on to that. It's not a binary thing. You
- 20 can't say that people say they care about privacy
- but they continue to use Facebook.
- You have to look deeper. You have to

- look at about how they're using Facebook, whether
- they're using the privacy controls, how they're
- engaging in those services, because if you look
- 4 deeper you see some pretty sophisticated choices
- that people are making in ways to protect their
- 6 privacy that's not apparent on the fact that, oh,
- you're using a social network, you must not care
- ⁸ about privacy.
- MS. WALD: I have a question. Between
- the two panels, the first panel and the second, I
- heard, I hope correctly, that there is some
- difference of opinion on a couple of things, or
- maybe slight.
- I think, Ms. Anton, you suggested in
- answer to a prior question of mine that you
- thought the government was indeed involved in
- trying to build privacy into the technological
- 18 aspects of some of the programs.
- On the other hand, earlier I think you
- said that in threat modeling very little privacy
- 21 considerations were going into that.
- Other people said that it wasn't

- inevitable that the government would keep
- ² collecting more and more information, but I think
- ³ I got that impression that maybe it seemed to be
- 4 going that way from Mr. Felten on the earlier
- ⁵ panel.
- So my question is basically, very
- ⁷ briefly, if there were one area of priority, if
- you were running the government's overall privacy
- ⁹ protection that you would suggest they
- concentrate on and could perhaps improve privacy
- 11 protection without endangering national security,
- what would it be? If you can do it very quickly.
- MS. ANTON: I think that we really need
- to work more on privacy standards and not privacy
- standards globally, and also that aren't rigged
- in some way to help some government or sector of
- industry. I think that's the number one
- 18 challenge right now.
- MS. WALD: Other people?
- MR. BEDOYA: Yeah, I would say it's
- 21 ending programs that involve the bulk collection
- of American's data.

- MS. WALD: I couldn't hear the end.
- MR. BEDOYA: Ending programs that
- involve the bulk collection of American's data.
- MS. WALD: Okay. Do you have in mind
- 5 any except 215?
- 6 MR. HINTZE: I didn't have the TSSI
- ⁷ clearance so I don't know.
- MR. DEMPSEY: Okay, Mr. Chairman?
- MS. WALD: Wait a minute, there was
- somebody wanted --
- MR. DEMPSEY: Oh, I'm sorry. Yes,
- 12 please.
- MR. NAHARI: One last thing, and I don't
- 14 know if this is the elephant in the room. One
- thing I would put as an item priority is our
- systems and the technology are very much built as
- one way. So I would introduce a notion of
- 18 revocation.
- So if something goes bad right now, if
- ²⁰ I'm releasing all of this information, there is
- 21 no way for a user, for a citizen to go ahead and
- 22 push a button somewhere and say revoke all the

- 1 rights that I gave to XYZ service providers and I
- want to go ahead and clear everything.
- 3 So defining what that revocation means,
- 4 what are the ramifications of that, and how to
- ⁵ crystallize it as a requirement for the industry
- 6 would go a long way for things that we could
- ⁷ build.
- 8 MS. WALD: That would go primarily to
- 9 industry, that wouldn't affect government. I
- mean if I gave the government some information
- under some program which I thought was going to
- benefit me and later on it turned out it was
- being used in a different way, would your
- 14 revocation principle apply there?
- MR. NAHARI: If I have the right to
- revoke whatever government had collected about me
- and I knew things that our government, in the
- possession of government and I was able to revoke
- that, perhaps that would be helpful.
- MS. WALD: Thank you.
- MR. DEMPSEY: So this concludes our
- second panel. It concludes our morning session.

- We will reconvene at 1:15 with a panel of
- ² government privacy officers.
- ³ (Off the record.)
- MR. MEDINE: Good afternoon. The
- ⁵ Privacy and Civil Liberties Oversight Board's
- 6 meeting on defining privacy will continue with
- our afternoon session with government panelists
- 8 moderated by a member, Beth Cook.
- 9 MS. COLLINS COOK: So welcome back to
- 10 folks who were here earlier, or welcome to those
- who were not here.
- Just quick one piece of housekeeping,
- what we've noticed this morning is make sure, and
- 14 Alex, this will be particularly relevant for you,
- make sure the microphone is actually the
- direction you are talking, so that even if you
- 17 pull it in front of you but then turn to talk to
- us, make sure the microphone is picking up. They
- were having problems this morning and we've all
- been gently reminded as well.
- All right, so this panel is about the
- 22 privacy interests identified and addressed by

- government privacy officials.
- Obviously in the counterterrorism
- 3 context defining and expressly articulating
- 4 individual privacy interests while balancing the
- 5 needs of national security is an extremely
- 6 challenging task.
- As we discussed a bit this morning,
- 8 widely accepted privacy frameworks like the Fair
- 9 Information Practice Principles or traditional
- 10 privacy impact assessments may very well be
- intentioned with the necessity to protect
- information regarding the operation of a
- 13 particular counterterrorism program.
- By the same token, some counterterrorism
- programs could be better served with greater
- transparency about what information is being
- collected, about the statutory authorities or the
- 18 authorities pursuant to which programs are being
- operated, and about what protections the
- government utilizes to minimize the negative
- impacts on individuals' privacy.
- So the panel that we have assembled

- today for this forum is, I think, uniquely
- ² situated to discuss these privacy issues that
- 3 arise in the context of federal counterterrism
- 4 programs.
- 5 These officials not only assess the
- 6 privacy impacts of a full spectrum of
- 7 counterterrorism programs they have also been
- pioneers, many of them, in the practice of
- ⁹ working proactively within the agencies to ensure
- privacy and civil liberties concerns are taken
- into consideration from the beginning of
- 12 programs.
- And if that were not enough of their
- duties, they also are learning to live with us
- 15 and work with us.
- Joining me today are three individuals.
- Unfortunatly DHS was not able to make anyone
- ¹⁸ available for this as it turned out.
- So we have three folks. They will have
- ten minutes, given that they have a little bit of
- extra time, few folks, but we will follow the
- 22 same basic framework.

- I will then ask a series of questions
- for a period of time and then invite my fellow
- 3 panelists to submit questions as well.
- So leading us off is Alex Joel who is
- 5 the Civil Liberties Protection Officer for the
- Office of the Director of National Intelligence.
- Do you actually fit that on one card?
- MR. JOEL: Yes, I do.
- 9 MS. COLLINS COOK: That's amazing.
- So in that capacity he leads the ODNI's
- 11 Civil Liberties and Privacy Office and he reports
- directly to the Director of National
- 13 Intelligence.
- Prior to joining the government, and I
- think this is also relevant based on our other
- panels, Alex served as the privacy, technology
- and e-commerce attorney for Marriott
- 18 International, where he helped establish and
- implement Marriott's global privacy compliance
- program, including the creation of Marriott's
- 21 first privacy officer position.
- So, Alex, did you want to kick us off?

- MR. JOEL: Yes, thank you. And I want
- 2 to thank the Board for --
- MS. COLLINS COOK: Oh, I'm sorry,
- 4 there's a stop light function going on here,
- green, good to go, yellow, start wrapping up,
- for red, stop, in the front row.
- 7 MR. JOEL: Okay. I want to thank the
- 8 Board for inviting us here to address the public
- ⁹ in this very important hearing.
- And as you said, the Board does work
- very closely with us. We feel that the Board's
- 12 role in providing both transparency and
- oversight, as well as advice to the intelligence
- community has been extremely valuable and is a
- critical part of how the intelligence community
- protects privacy and civil liberties.
- So I want to thank the Board for holding
- this hearing and for the Board's very diligent
- and careful efforts to exercise their statutory
- functions, which I think have been critically
- important.
- This topic is, of course, one that

- $^{
 m 1}$ consumes all of us, not specifically how to
- define privacy, but how to apply protections
- 3 required to protect privacy in the context of our
- 4 activities and in particular in the context of
- ⁵ counterterrorism activities.
- I'd like to just get to what I think of
- ⁷ as the heart of the matter from an intelligence
- 8 community perspective in any event, which is that
- ⁹ we operate by necessity within a sphere of
- secrecy.
- We have to be able to maintain secrets
- in order to be effective. The more publicly
- transparant an intelligence service is, the more
- it informs adversaries of how the agencies are
- collecting information and the better able those
- ¹⁶ adversaries are to avoid detection.
- So as I've said in the past, a fully
- transparent intelligence service is by definition
- an ineffective one.
- The key for us then is how within the
- sphere of necessary secrecy do you make sure that
- the intelligence agencies are acting

- appropriately, lawfully, and in a way that
- protects people's privacy and civil liberties
- 3 consistent with the values of the nation.
- In the past what we have done, as you
- 5 know, is focused on ensuring that we are
- 6 providing full transparency to our oversight
- ⁷ entities. And our oversight system is something
- 8 that I would like to characterize as a system of
- 9 many layers with many players.
- We have not only within each agency,
- offices of general counsel and offices of
- inspectors general, as well as newly created
- 13 privacy and civil liberties offices, but outside
- of the agency we have entities like the
- Department of Justice, which is responsible on a
- government-wide basis for exercising some of
- these authorities and oversight controls.
- We have of course newly created entities
- 19 like the Privacy and Civil Liberties Oversight
- Board, perhaps not that new anymore, which again
- is designed to make sure that there is a secure
- 22 place for information to be disclosed and

- discussed so that the oversight institutions are
- ² satisfied that the activities being conducted are
- ³ proper ones.
- Then of course we have Congress and the
- ⁵ judiciary, both of which exercise robust
- oversight. And I would mention that, for
- ⁷ example, the congressional oversight committees,
- 8 which were established particularly after the
- 9 Church Committee hearings in the 1970s to provide
- this granular level of oversight over
- intelligence activities, has been very effective
- in my view in providing careful oversight of what
- 13 we do.
- So that's sort of the oversight part of
- the equation. I think what we have now more
- fully realized is the need to enhance
- transparency.
- So if you think of it, I mean I was just
- thinking about this before I started talking,
- which is always dangerous, but if you think of it
- 21 as operating within a sphere of secrecy, one way
- is to make sure that the mechanisms, the rules

- and oversight structure within that sphere are
- 2 robust enough to make sure that privacy interests
- and civil liberties interests are being
- ⁴ adequately protected.
- And then there's the other way of
- 6 approaching this, which we're also focusing on
- doing, which is reducing that sphere.
- In other words, providing greater
- ⁹ transparency into what goes on inside the
- intelligence agencies so that the public at large
- 11 can get reassurance and can also provide input
- 12 and feedback into how we conduct these
- ¹³ activities.
- I think if I could just continue along
- this theme, there are two aspects in particular
- of what goes on to regulate our activities that I
- 17 think is of interest. One is the rules that we
- 18 follow, and the other is the oversight framwork
- and mechanisms designed to make sure we're
- 20 following those rules.
- So I think on the former, what are the
- rules that we follow? We can and should provide

- greater transperancy, but a lot of those rules
- ² are now currently being debated and discussed,
- and you can think of some of the reform
- 4 mechanisms as attempts to modify those rules.
- 5 So you have the activity going on in
- 6 Congress, for example, the USA Freedom Act and
- ⁷ similar legislative initiatives.
- 8 You have as part of that also the
- 9 proposal to create an advocate of some kind, an
- adversarial mechanism for the Foreign
- 11 Intelligence Surveillance Court.
- Here again in my view is an attempt to
- influence or affect what are the rules that the
- intelligence agencies are expected to follow.
- And then a different part of that
- question is what oversight mechanisms, what
- 17 assurances do we have that the agencies are, in
- 18 fact, following those rules.
- And you're part of that. I've already
- mentioned the congressional committees, the
- Foreign Intelligence Surveillance Court, and then
- 22 all the layers within the executive branch itself

- at the intelligence community and the Department
- ² of Justice level.
- So I think, I hope that the public
- 4 discussion has been shifting a bit from whether
- or not we're following the rules. I think what I
- 6 perceived in the public discussion is a greater
- acceptance that we are in fact trying our best to
- 8 follow the rules. We're not perfect and we make
- 9 mistakes, but we're trying to follow those rules
- as best as we can.
- And now the discussion has been shifting
- to, well, what should those rules be? What are
- the rules, and what should those rules be?
- I think we can and must provide greater
- transparency into both sides of that equation,
- and we're working on that.
- I would also say that another thing that
- 18 I know the Board has been pursuing which is the
- recommendation that the Board made in the 702
- report regarding efficacy, you know, to what
- 21 extent are the counterterrorism programs and
- measures effective and to what extent do they

- 1 provide value is a key part, in my view, of the
- ² transparency equation as well.
- We have to figure out ways to identify
- 4 the specific value associated with particular
- 5 programs and activities, and then be more
- transparent about that so that the American
- people can render a judgment, as well as everyone
- 8 else, on the need or desirability for a
- ⁹ particular kind of program.
- 10 It is very difficult to do all this
- 11 stuff and still maintain secrets. The
- intelligence communitity is not built for
- 13 transparency. I've said this before, it's built
- 14 for exactly the opposite, of course.
- We train, provide policies and systems
- and reminders to our workforce of the importance
- of maintaining secrets, you know, maintaining
- 18 secret the sources and methods that the
- intelligence community uses to carry out its
- 20 activities. And this is vital. I mean we have
- to do that and we're reminded of that need all
- 22 the time.

- But at at the same time, we have to find
- ways to enhance transparency. It's going to
- involve some changes in culture, training, a look
- at policies and processes within the intellgence
- 5 community and I know that you may want to ask
- 6 questions about that, so I look forward to that
- ⁷ discussion.
- 8 So thank you again. I appreciate it.
- 9 MS. COLLINS COOK: So turning now to
- 10 Erika Brown Lee, she is the Chief Privacy and
- 11 Civil Liberties Officer of the Department of
- Justice. In that capacity she is the principal
- 13 advisor to the Attorney General on privacy and
- 14 civial liberties matters affecting the
- department's missions and operations.
- And as part of the Office of Deputy
- 17 Attorney General, Ms. Brown Lee oversees the
- department's privacy and civil liberties programs
- ¹⁹ and initiatives implemented by department
- 20 components and component privacy and civil
- 21 liberties officials.
- She also heads the Office of Privacy and

- 1 Civil Liberties, which reviews and evaluates
- department programs and initiatives, and provides
- department-wide legal advice and guidance to
- 4 ensure compliance with applicable privacy laws
- 5 and policies, including the Privacy Act. Thank
- 6 you for coming.
- MS. BROWN LEE: Thank you, and thank you
- 8 to the Board for inviting me here to talk about
- ⁹ what is a very important topic.
- 10 You asked about private sector
- experience and other government experience, I
- 12 also come from the Federal Trade Commission,
- which in particular the Division of Privacy and
- 14 Identity Protection, which of course the Federal
- 15 Trade Commission has a very different orientation
- toward the commercial side of privacy, but
- 17 nonetheless an important perspective and an
- interesting one to bring to this position.
- But counterterrorism is a significant
- 20 part of the department's mission. Since my
- 21 colleagues on the dias today will be talking from
- more of an intelligence lens, I thought I would

- orient my remarks more toward the department's
- efforts to fight terrorism from within the
- 3 criminal law enforcement context.
- 4 The department has an elaborate
- 5 architecture that protects privacy in our
- 6 counterterrorism work, and since I only have a
- few minutes I'll focus on the lead agency in
- 8 those efforts, which is the FBI and focus in a
- 9 little bit more on the efforts with their
- 10 counterterrorism activities.
- But stepping back for a minute, of
- course as we know after 9/11, it was recognized
- that in order to address the current threat
- environment, FBI's functions needed to be
- expanded, but it was not intended that the
- expansions would come at a cost of civil
- 17 liberties.
- So in 2008, the department issued the
- 19 Attorney General Guidelines for Domestic FBI
- Operations, the AGG-DOM, and later that year
- issued the DIOG, or the Domestic Investigations
- Operations Manuel. And combined, those two

- documents provide significant guidance for FBI
- ² activities.
- But what I wanted to talk about, and I
- know I don't have enough time to get too far into
- the weeds, is just to explain how privacy is sort
- of embedded throughout the stages of an
- investigation, from the initial phase throughout
- 8 the process.
- And so, for example, one of the key
- tenants of both documents is the least intrusive
- 11 method. So in other words, in any activity that
- the FBI engages, that's the baseline.
- But of course within the
- 14 counterterrorism context, it's got to be
- calibrated against the threat to national
- security, in which case more intrusive methods
- would be used.
- But in terms of a little bit more detail
- 19 from an operational context, when an FBI conducts
- an assessesment, for example, which necessarily,
- well not necesssarily, but oftentimes is
- proactive, that would involve, doesn't require a

- 1 factual predication, but it does require a
- ² clearly defined objective.
- And the least intrusive methods in that
- 4 context would be even starting with publically
- ⁵ available information, to voluntarily provided
- information, in that perspective.
- And then moving up from there with
- 8 regard to predicated investigations, which of
- ⁹ course implies by title, there requires a factual
- predication to open that investigation, but that
- has to have supervisory approval.
- 12 And both types of investigative
- 13 activities, whether it's assessments or
- 14 predicated investigations require or are, I
- should say, subject to oversight.
- Alex mentioned DOJ oversight on the
- intelligence side, but also on the law
- enforcement side for counterterrorism, the
- department's National Security Division has
- oversight authority for those kinds of
- ²¹ activities.
- Now Beth mentioned and asked us to talk

- about or think about how the FIPS apply, if
- you're looking for the acromyn, there's lots of
- them in the documents, but it's not actually in
- 4 the AGG-DOM or the DIOG. However, they are
- ⁵ embedded throughout really, the princples.
- If you think about, even from a
- ⁷ transparency perspective, right, all that I'm
- 8 discussing with regard to the DIOG, all 700 pages
- ⁹ of it for a little light reading, for anyone
- who's interested it's on the web with certain
- 11 redactions.
- But also we have privacy impact
- 13 assessments that are available. And one that I
- wanted to just mention in particular regards the
- eGuardian system because that is a specific
- system or incident reporting system that is
- designed as a platform to share terrorism-related
- information across law enforcement, you know,
- 19 federal, state, local, tribal, territorial
- ²⁰ jurisdictions.
- So eGaurdian, I don't have time to go
- into much detail about it, but it has an entire

- architecture of privacy protections governing how
- information comes into eGuardian, how it's shared
- across those entities, how it's stored and how
- 4 it's retained.
- 5 Individual participation as a FIPS
- ⁶ principle, obviously that's more of a challenge
- in a law enforcement context. It's not realistic
- 8 to be able to obtain individual consent in order
- ⁹ to pursue criminal investigations.
- But nonetheless, the Privacy Act
- 11 provides some measure of review in the sense that
- if access or amendment to records is denied,
- there is judicial review of an agency's decision,
- 14 and subject to court order, records may be
- amended or access may be granted.
- On the minimization side, I mentioned
- the least intrusive means already with the DIOG.
- There's also a prescriptive measure in the DIOG
- with regard to evidence collected, that if the
- evidence collected through an assessment or
- through a predicated investigation has no
- forseeable future evidentiary or intelligence

- 1 value, it should be returned and destroyed, and
- then marked in the file in term of the
- disposition of that piece of evidence.
- 4 Otherwise, information is retained
- 5 according to the schedule set by NARA, the
- 6 National Archives and Records Administration, and
- ⁷ approved, through which the Department of Justice
- 8 would seek approval for.
- With regard to use, I think that's also
- 10 a challenge. On the criminal side of course
- willful disclosures of protected information
- under the Privacy Act are not something that any
- 13 agency can exempt themselves from.
- And to the extent that information is
- 15 released that's not subject to a routine use or
- other permitted disclosure, and of course, you
- know, routine uses are subject to a compatibility
- 18 standard that tracks the FIPS language.
- 19 If the information is disclosed or even
- shared in violaton of that, that's potentially a
- wrongful disclosure subject to not only civil
- damages but criminal penalties.

- And then in terms of accountability, I
- ² mentioned oversight from the National Security
- Division, but also the FBI has the National
- 4 Security Law Branch, which conducts national
- ⁵ security reviews.
- And that's a significant review
- ⁷ process in that they go out to all of the field
- 8 offices and review the investigative activities I
- 9 mentioned, the assessments, the predicated
- investigations and look to see whether, in fact,
- superviseory approval was obtained, whether, in
- 12 fact, there was a clearly defined objective for
- any assessment, and it's written up into a
- 14 report.
- 15 That report actually comes through FBI
- channels of course, but then also comes for
- 17 review by the Chief Privacy and Civil Liberties
- Officer. And I look at those, obviously, through
- ¹⁹ a privacy and civil liberties lens.
- So as Alex was mentioning, there are
- lots of layers that are applicable.
- I know I don't have much time remaining.

- 1 But in conclusion, I guess I would just like to
- leave you with a couple of take-aways.
- One is that FIPS, quite to the contrary
- 4 of certain statements is not dead, it's just
- ⁵ embedded.
- And I would also say that the processes
- ⁷ can always be improved. Certainly I work with
- 8 the component, each component. There are over 40
- 9 components in DOJ, but each component has a
- 10 Senior Component Official for privacy and I host
- 11 regular meetings.
- In fact, we're having a privacy forum
- 13 next week that will cover privacy-related
- 14 activities focusing on law enforcement, but other
- components as well, activities, common privacy
- issues across components. It is internal though
- so none of you are actually invited unless you
- happen to get a job by Monday at the DOJ.
- But that's also something that is a way
- to improve. And I would also say that while
- 21 privacy impact assessments are very important and
- 22 a critical part of our program because they're

- sort of this tangible proof that we actually
- evaluate privacy, that we mitigate the risks,
- 3 that we take into account security and
- 4 accountability, they really only form a part of
- the architecture for the Department of Justice's
- ⁶ privacy program.
- So, and I welcome your comments.
- MS. COLLINS COOK: Thank you, Erika, for
- ⁹ a nice education about the FBI's operations, the
- ¹⁰ FBI in particular.
- So Becky Richards is the National
- 12 Security Agency's Civil Liberties and Privacy
- 13 Officer. In this, I think, relatively new role,
- 14 I think it's fair to say, she provides expert
- advice to the Director of NSA on all issues
- 16 pertaining to privacy and civil liberties
- 17 protections, and she conducts oversight of NSA's
- 18 civil liberties and privacy-related activities.
- She also develops measures, which I hope
- she will talk about, to further strengthen NSA's
- 21 privacy protections.
- Prior to joining the National Security

- 1 Agency, she worked as the Senior Director for
- ² Privacy Compliance at the Department of Homeland
- 3 Security.
- MS. RICHARDS: Thank you, and thank you
- ⁵ for hosting us. I am very honored to have been
- 6 selected to be the first NSA's Civil Liberties
- ⁷ and Privacy Officer.
- This is an exciting time to be a member
- ⁹ of the civil liberties and privacy community.
- Our community is growing and evolving and will
- 11 help inform the debate as the nation reshapes its
- expectations for and limitations on the
- intelligence community activities.
- 14 Changes in the nature of the threat to
- our national security, alongside rapid advances
- of technology, as was discussed earlier, make my
- job both interesting and challenging.
- 18 Technology provides us with both
- opportunities and challenges, but ultimately we
- must guide and shape its use to ensure the
- fundamental rights we hold dear as a nation are
- ²² maintained.

- Today I'd like to take a little time to
- describe NSA's civil liberties and privacy
- programs, both in the past, present, and a few
- 4 thoughts on the future.
- 5 Part of the NSA's mission is to obtain
- foreign intelligence worth knowing derived from
- ⁷ foreign communications in response to
- 8 requirements and priorities validated and levied
- ⁹ upon us by the executive branch.
- One such priority is counterterrorism,
- but there are other threats to the nation, such
- 12 as the spread of nuclear, chemical or biological
- weapons, or cyberattacks.
- NSA also works directly with and
- supports our troops and allies by providing
- foreign intelligence for military operations
- ¹⁷ abroad.
- As we consider NSA's civil liberties and
- privacy programs over the past 62 years, it's
- important to think about how the threat,
- technological and sociatial landscape in which
- NSA conducts itself signant mission has changed.

1 First, the threat has changed. 2 previously operated in a cold war era when the 3 focus of collection for foreign intelligence was directed at nation states, structured military 5 units, and foreign intelligence services. 6 While threats remain from nation states, 7 they now also come from non-state actors, which 8 require NSA to look at more, smaller and 9 decentralized targets to protect the nation. 10 The technology has changed. NSA again 11 previously operated in an environment where the 12 communications between foreign intelligence 13 targets were frequently conducted over isolated, 14 government-owned and operated communication 15 channels and equipment. 16 Now foreign target communications are 17 interspersed with ordinary commerical and 18 personal communications. 19 Additionally, the sheer volume and 20 ability to analyze and manipulate big data, which 21 has occurred as a result of significant advances 22 in information technology, can expose information

- of a personal nature that may not have been
- ² previosly discoverable and may not be of any
- ³ interest.
- 4 Third, how society thinks about civil
- liberties and privacy has changed. We've come a
- 6 long and positive way in thinking about what
- ought to be private.
- Personally identifiable information was
- 9 not a mainstream issue 25 years ago. For
- example, Social Security numbers were routinely
- 11 put on student ID cards and there was no thought
- of HIPAA.
- So with that I'd like to give a little
- 14 historical perspective. NSA's civil liberties
- and privacy protections have historically been
- driven primarily by the Fourth Amendment
- analysis, which is also reflected in NSA's
- authorities, Executive Order 12333 Foreign
- 19 Intelligence Surveillance Act, or FISA.
- This analysis framed NSA's protection
- 21 program by asking where and how the data was
- collected, i.e., usually overseas, and the status

- of the individual or entity being targeted, i.e.,
- 2 is it a U.S. person or not.
- NSA has consistentily conducted
- 4 extensive legal analysis as it considers new
- 5 types of collection answering these types of
- questions. It has built a strong compliance
- 7 program based on these, with compliance
- 8 activities embedded in our technologies and
- 9 systems.
- As I have learned more about NSA and its
- 11 compliance regime, it became clear while this is
- certainly one way to address privacy concerns, it
- is somewhat different from how privacy concerns
- ¹⁴ are addressed outside of NSA.
- Over the last 15 years Congress has
- passed a variety of laws to protect privacy in
- other parts of the government and the commercial
- 18 sector. These policies and laws focus more on
- the nature and content of the data and how it is
- used, not where it was collected or the
- 21 citizenship of the individual.
- I believe we have an opportunity to

- bring together NSA's current civil liberties and
- ² privacy analysis with a broader approach to
- ³ privacy and civil liberties.
- 4 This new approach also supports the
- 5 President's PPD-28 mandate to recognize that our
- 6 signals intelligence activities must take into
- ⁷ account that all persons should be treated with
- 8 diginity and respect, regardless of their
- 9 nationality and wherever they might reside, and
- that all person's have legitimate privacy
- interests in handling their personal information.
- To address a broader set of civil
- 13 liberties and privacy interests, I'm testing a
- 14 civil liberties and privacy assessment process
- that expands NSA's views to include
- 16 considerations of frameworks the private sector
- and nonintelligence elements of the government
- use to assess civil liberties and privacy.
- For example, for the first time in its
- history, NSA is using the Fair Information
- 21 Practices Principles, or FIPS, as a framework for
- considering civil liberties and privacy risks.

- 1 The FIPS are one framework through which
- organizations can analyze the protections they
- 3 have in place for personal information.
- 4 While traditional NSA civil liberties
- 5 and privacy questions center on citizenship and
- 6 location of foreign intelligence targets, as well
- as collection techniques, FIPS related questions
- 8 boil down to follow the data.
- 9 Data-centric perspectives mean privacy
- officials ask a different set of questions. What
- is the data being collected and how will it be
- 12 used?
- 13 As such, we've designed an initial
- 14 standarized template and during the next year
- we'll refine the questions and process to ensure
- we're building a repeatable, meaningful and
- helpful process to identify and mitigate civil
- 18 liberties and privacy risks.
- A critical part of the civil liberties
- and privacy assessment process is to make sure
- we're not merely checking off boxes, but
- fundamentally weighing the risks associated with

- the activity to form a holistic value
- ² proposition.
- In essence, we're asking should NSA
- 4 conduct a given activity, given its civil
- ⁵ liberties and privacy risks?
- As part of the assessment process NSA is
- documenting both standard protections, such as
- 8 minimization and control on who has access, as
- 9 well as any specialized tools designed to protect
- civil liberties and privacy.
- Much like privacy analysis performed in
- the private sector and other parts of the
- government, we're using the FIPS as the basis for
- analyzing what existing protections are in place.
- As we look to the future, I'd like to
- spend a little bit of time talking about blending
- the art and science of privacy.
- Historically privacy tends to be a bit
- of an art form. Several of us stand around and
- think about how we're going to do the analysis.
- This can be difficult when we're beginning to
- think about big data and the complexity that was

- being discussed this morning.
- NSA is fundamentally a technology-
- ³ centric organization. We have and will continue
- to contribute to advancing the discussion and
- ⁵ research of protecting civil liberties and
- ⁶ privacy.
- 7 Today the science of privacy has made
- 8 notable strides that include developing
- ⁹ technology and tools that promote privacy, such
- as unique encryption capablilities, digital
- rights management and trustworthy computing.
- Great work in private sector and
- 13 academeia is also being developed on coding
- 14 privacy policies, such that technology supports
- ¹⁵ all specific uses.
- But civil liberties and privacy
- 17 protections need to blend the art and science of
- privacy if we're going to harness the potential
- of technology and incorporate our core values as
- 20 a nation into this era of big data.
- So despite significant progress in
- 22 privacy technology, basic privacy of principles

- found in a strong scientific basis, have largely
- ² proven elusive.
- If we can better understand what
- 4 constitues personal information and how such
- information is used, we believe it will be
- 6 possible to determine whether we can develop more
- ⁷ practical approaches to evaluating the inherent
- 8 risk of privacy to the individual.
- To that end, our initial thoughts are to
- develop five sequential building blocks and to
- introduce the concept of some very difficult math
- into what is otherwise a very nice liberal arts
- discussion of privacy.
- The first one is to catagorize personal
- information. We would like to determine if it's
- possible to identify and catagorize different
- types of personal information and what that risk
- is to privacy.
- Now we've heard different discussions
- today, but we want to push folks to think about
- is certain types of data more risky to privacy,
- say likehealth data, than other information, say

- 1 your address, and can we think about those risks.
- If we can do that, then next we would
- like to determine if it is possible to identify
- 4 and catagorize different types of use.
- If we take both of these together, it's
- 6 possible to develop a catagorization of both
- 7 personal information and uses of the personal
- information, it should be possible to develop a
- 9 scientific process to assess risk.
- This process could evaluate the risk of
- the use of individual types of personal
- information for different uses, as well as
- aggregated uses of personal information.
- Now with these three building blocks
- being more of the scientific aspect, I would now
- suggest we would move to an art form that looks
- 17 at how we build that to identify what needs to
- have additional privacy impact analysis
- 19 conducted so that we're looking at that across
- 20 the board.
- With all four of these together then we
- would look to see if we could build a responsible

- use framwork that holds data collectors and users
- ² accountable for how they manage data and any harm
- ³ it causes.
- 4 Building a technical means based on
- ⁵ principled scientific methodologies to support
- the identification of civil liberties and privacy
- ⁷ risks can help us better protect civil liberties
- 8 and privacy in a fluid world of big data.
- 9 Success is dependent upon input from a
- variety of disciplines ranging from
- technologists, social scientists, privacy and
- 12 civil liberties experts, ethicists, attorneys and
- computer scientists, to name a few.
- We would welcome the opportunity to
- discuss this in more detail and greater technical
- depths at a later date.
- With that, I thank you for the
- opportunity and I'm happy to answer what I'm sure
- ¹⁹ are a couple of questions.
- MS. COLLINS COOK: Thank you all for
- your opening remarks.
- Becky, I wanted to stick with you for

- just a second. When we go and meet with y'all
- and when we talk to y'all, there is frequently
- 3 someone from the general counsel's office,
- 4 someone from the compliance office, someone from
- ⁵ your office.
- What are you doing that is different
- ⁷ than the general counsel's office and a
- 8 compliance shop?
- 9 MS. RICHARDS: That's a great question.
- 10 So the civil liberties and privacy office at NSA
- is the focal point for questions surrounding
- civil liberties and privacy, and it's been
- brought to a senior leadership position at NSA in
- order to focus on those efforts.
- So generally speaking, our general
- 16 counsel will answer the legal question, is this
- 17 legal permissable? And they will often then work
- with compliance for, what are the rules?
- But we haven't had a person asking some
- of these more difficult questions of, should we
- 21 be doing this?
- Now frequently our oversight folks,

- whether it's ODNI and DOJ, were playing that
- role. And so I don't want you to take away the
- idea that those questions weren't asked.
- But it's really important to have that
- 5 type of a role inside the building where you are
- 6 working with the operators and the technologists
- ⁷ and can spend a great deal of time understanding
- 8 what we're trying to do and bring to bear those
- ⁹ questions.
- MS. COLLINS COOK: Erika, a similar
- question for you. FBI, for example, has its own
- 12 privacy officer, has its own general counsel, has
- its own compliance shop.
- What is your relationship and what is
- your ability to provide recommendations or to
- actually impose requirements on the FBI?
- MS. BROWN LEE: So also a very
- interesting question. My role and position is
- department-wide, so of course I have oversight
- over the compliance for DOJ as a whole.
- Each component, as I mentioned, has a
- senior official for privacy, but in addition has

- general counsel's office that has significant
- footprints in privacy. So at FBI they have their
- 3 privacy and civil liberties unit that's headed by
- ⁴ a chief.
- I work quite significantly with that
- 6 person in that office to specifically address
- ⁷ compliance issues, to specifically address
- 8 privacy initatives that I feel are important for
- ⁹ the bureau to consider.
- Ultimately it is somewhat of a reporting
- 11 structure. In other words, if there is a
- 12 recommendation, or a particular policy or
- 13 statutory obligation, FBI has the responsibility
- 14 to comply.
- But part of what my job is, is to
- advocate and to make sure that that is occurring
- on a regular basis and that looking for ways that
- 18 I can improve the process, looking for ways, for
- example, I talked about privacy impact
- 20 assessments. Some of that is, if you look at the
- 21 E-Government Act, it's written fairly broadly.
- I take, you know, a particularly broad

- 1 view of what I think should have assessments as
- ² part of compliance there. And so that's what I
- work in particular with the FBI on.
- 4 MS. COLLINS COOK: So Alex, a related
- but different question for you. How do you
- ensure that you have access, do you ensure that
- you have access to what various agencies are
- 8 doing, or do you find yourself periodically
- 9 reading about new programs, alleged new programs
- on the front page of the New York Times?
- MR. JOEL: I'm surprised by that
- 12 question. Information sharing is perfect
- everywhere in government.
- MS. COLLINS COOK: I'm also seeking free
- advice because obviously one of our biggest
- challenges is going to be knowing what the
- agencies are doing. You can't conduct oversight
- of something you don't know is happening.
- MR. JOEL: Right. I think that it's a
- 20 major challenge for all of us. I know that, as
- you said, it's something that you're focused on.
- I know that it's a challenge for everybody.

1 It's a matter of, first of all, 2 understanding the information flows within your 3 own agency and trying to put in place markers for where it's important for you to be consulted. 5 The main way that I have just 6 practically done it, since I've been doing this for about a decade now and when I first started, 8 you know, it was just me and then we built a 9 small staff over time, has been to form the 10 trusted relationships inside the intelligence 11 community and to make sure that the people that 12 I'm working with and that are in positions of 13 influence and authority to make decisions on 14 programs and activities, understand the 15 importance of consulting with civil liberties and 16 privacy professional. 17 In my own personal experience working 18 within the intelligence communitity has been that 19 when I first joined I was very pleasantly 20 surprised that people were so focused on 21 compliance and protecting privacy and civil 22 liberties, doing the right thing, following the

- 1 right directives, and even when they might feel
- legally permitted to do something, they still
- gave voice to their own doubts as to whether they
- 4 should be doing it.
- 5 And so I did not personally experience
- an uphill battle in trying to pursuade
- ⁷ intelligence officers, hey, it's important for
- you to pay attention to civil liberties and
- ⁹ privacy.
- In fact, it was sort of the opposite
- where many people felt that they were already
- doing that, and that it was their job to focus on
- 13 that.
- For example, you mentioned Office of
- 15 General Counsel. I was at an office of general
- counsel before coming to this job and we
- 17 certainly felt when I was there that that was
- 18 part of our job. We needed to look out for
- 19 privacy and civil liberties, and not just what
- the law allowed, but what was the underlying
- intent and what should we be doing in that light.
- So I certainly didn't want to take away

- that sense of responsibility from anybody inside
- ² the intelligence community.
- My approach had always been, it's all of
- our jobs, it's part of our oath to support and
- ⁵ defend the Constitution. There are offices that
- 6 are particularly focused on that, Office of
- General Counsel, Office of Inspector General.
- 8 There are intelligence oversight offices, as you
- ⁹ guys have learned, that are.
- Now we're creating these civil liberties
- 11 and privacy offices and I do think we add value
- because I think it is our full-time job to focus
- on civil liberties and privacy, so we bring
- 14 focus, we bring an external perspective, and we
- 15 have specific expertise, and training and
- experience that we can bring to bear, and then we
- can become a voice, as Erika said, an internal
- 18 advocate for civil liberties and privacy.
- But I mean I think different agencies
- will find different ways of doing it. The ODNI
- is a fairly small organization, and the ODNI
- 22 itself has mechanisms for understanding what's

- going on across the intelligence community. So
- when a particular program or activity bubbles up
- 3 to the point of a decision, either it comes
- 4 automatically through my office or somebody will
- understand that I need to see it and route it to
- 6 me.
- MS. COLLINS COOK: So a follow-up,
- 9 particularly to you, Alex, and Erika, both of you
- 9 have fairly small staffs considering the breadth
- of your responsibilities, and we talked a lot
- this morning about the increasing technological
- 12 complexity of what you are assessing.
- Do you have the technological resources
- 14 to understand what systems are actually doing?
- ¹⁵ And I think that is both in terms of assessing on
- the front-end whether systems or programs should
- go live, or to the extent that there are
- 18 restrictions, for example, if the FISA Court puts
- 19 a restriction in place on a particular program,
- ensuring that those restricions are actually
- ²¹ functioning.
- MS. BROWN LEE: So I think that's a good

- point. So but as I mentioned earlier, oversight,
- there are sort of a variety of roles in the
- department that have oversight, particularly with
- 4 regard to counterterrorism.
- 5 But my office is fairly small in the
- 6 sense that given the large footprint of the
- ⁷ Department of Justice, but they work incredibly
- 8 hard and diligently with all of the components to
- 9 ensure compliance.
- We rely quite a bit on internal
- 11 component work that is done to produce
- information about what the privacy compliance is,
- 13 and then also with regard to auditing and making
- 14 sure that the privacy activities are actually
- ¹⁵ effective.
- But I would also say that some of the
- oversight, just to sort of again stress that,
- some of the oversight isn't just through my
- office, it's National Security Division, and FBI
- 20 also has their branch, so we work very
- 21 collaboratively.
- 22 And like Alex, I have found that within

- 1 the department there are a lot of people who care
- very deeply about these issues. It's not
- 3 specifically in a privacy role as a title, but
- 4 they have oversight and I think meaningful
- insight as to how the activities should consider
- and be consistent with privacy initiatives.
- But, you know, it is something that I
- 8 take into account and that's part of the reason
- ⁹ why we have these internal conferences and
- whatnot that I'm trying to do to build upon that.
- MS. COLLINS COOK: And Alex, what do you
- do to make sure, the old adage is trust but
- verify, what do you do to make sure you actually
- understand the programs and the systems?
- MR. JOEL: Right. So it's a variety of
- things. One is, although I am not personally a
- technologist, I have been dealing with technology
- law, and legal issues and privacy issues
- associated with technology for much of my
- 20 professional career.
- So when I was at Marriott, I was the
- privacy, commerce and IT lawyer there. And then

- before that I was at a law firm in downtown D.C.
- focused on large scale technology transactions.
- That doesn't make me a specialist in
- 4 technology, but it does enable me to ask the
- ⁵ right questions and make sure that the
- information is explained to me appropriately.
- I don't have the staff resources to
- 8 engage a full-time technologist. I think that
- 9 would be helpful. I do think that you have to be
- 10 a little bit careful with that because what you
- 11 really want in that sense is a technology
- 12 generalist.
- There are so many different aspects to
- to technology, as you know. I mean that's just a
- word that almost lacks meaning these days because
- we use it so frequently.
- But what NSA does for one particular
- 18 type of activity will differ significantly from
- what FBI does, will differ significantly from
- what all agencies do in terms of database
- management.
- So you've got database issues, you have

- surveillance technologies, understanding
- ² communications technologies, understanding all
- 3 kinds of different aspects to that issue.
- 4 And then of course the engineers and
- technologists, as we know, speak a different
- 6 language from lawyers and so sometimes it's hard
- ⁷ for everyone to speak to each other.
- So what I have been doing is making sure
- ⁹ that the information is clearly presented, that I
- see the documentation, that I personally
- understand it, that I trust the people who are
- 12 providing me that information are giving me a
- 13 complete picture, and then we also leverage
- technical experts in the particular field that we
- have access to within ODNI or through the agency.
- 16 So if something comes up that we don't quite
- understand, we can reach out to somebody to have
- them help us understand it.
- I think with a larger staff I would try
- to have more full-time technical expertise.
- MS. COLLINS COOK: Becky, you had
- mentioned that you've got a couple of pilot

- experiments going and you mentioned also new
- technologies that may or may not be available.
- How are you working with the private
- 4 sector to leverage what great thinking is going
- on, and is privacy a part of the procurement
- 6 process, for example? And has consideration been
- ⁷ given to that, that if we really want privacy to
- be from the ground up, should it be one of the
- 9 procurement factors?
- MS. RICHARDS: I'll start with the
- 11 procurement. We actually started with the
- theory on procurement because in part that's how
- we were doing things at DHS.
- But it turns out NSA is a technology
- company that has a huge research portion of it
- and it also has a huge technology division. So
- it's two different parts.
- So I actually have a technical director
- on staff who's here, Dave Marcos, and he and I
- have been working through sort of how do we think
- about the tech, how do we look at both what's out
- in the world, and so we're actually working with

- several different groups within NSA to do an
- initial review of what is out there right now.
- And they're conducting that right now so
- we can get a sense of both from a policy and a
- technology perspective what's going on, as
- opposed to just things that we may know just, you
- ⁷ know, from knowing different people, whether it's
- 8 activities going on at MIT or Carnegie Mellon,
- 9 you know, to make sure we had a broad breadth of
- understanding of what was the type of research
- 11 going on.
- So they're doing that. We're working on
- that right now, and then we're working with our
- 14 research folks and trying to just leverage all of
- 15 those things.
- The procurment process is not really
- helping this happen best at NSA. And I think
- that that's, you know, each agency has its own
- 19 culture and its own aspects. And so a lot of
- what I've been doing is taking the learning and
- sort of shifting it to make sure that building
- the program within NSA works for how NSA works.

- 1 And so that means that our privacy
- 2 program is going to look a little bit different
- than FBI's or others. But it's based on sort of
- 4 how the organization functions and where those
- being made. So we're working
- 6 through that.
- But it turns out procurement really
- 8 isn't really isn't quite the right place. So
- ⁹ we're looking through in terms of both the
- technology, and the research director and others
- to make sure we understand where those touch
- points are. And that's a lot of why we're beta
- testing the processes.
- MS. COLLINS COOK: So I think I have
- time for one last question before I turn it over
- to my fellow Board members.
- 17 Alex, this one's for you. You
- explicitly pointed to congressional oversight as
- one of the things that the American people should
- be aware of, that this happening, it's robust,
- it's real.
- A previous panelist pointed out that

- there is potentially one significant flaw or
- ² challange with congressional oversight, and
- 3 that's the lack of cleared staff.
- What has your perception been? Has
- 5 Congress struck the right -- yes, I'm going to a
- 6 ask you to opine on Congress -- whether
- 7 consideration should be given to broadening the
- 8 range of individuals?
- I think there's some comfort level with,
- 10 I think someone called it delegated oversight
- within the Congress. But when some significant
- majority of decision-makers in a representative
- democracy don't have cleared staff, how is the
- oversight nonetheless sufficiently robust?
- MR. JOEL: So the intelligence oversight
- 16 committees have very substantial cleared staff.
- 17 And they of course have secure compartmented
- information. We have SCIFs in which to review
- all the classified information. And we have
- many, many meetings, briefings and reports with
- our oversight committees.
- I guess my first response as a matter of

- 1 principle, yes, Congress should have the degree
- of staff cleared it needs in order to assist it
- 3 to perform its oversight functions.
- 4 I think that intelligence community
- 5 assumption had been that by clearing the staff of
- the oversight committees that that function was
- ⁷ being fulfilled.
- I think some staff members are also
- 9 cleared from some of the other committees. I
- don't have all of that information in front of me
- but I believe judicary has cleared staffers,
- 12 etcetera.
- Whether or not that's enough staff to be
- cleared, I don't know. I think Congress, from my
- personal perspective, it would be helpful if
- 16 Congress figured out for itself which committees
- 17 are performing which function and which staff
- members need to be cleared in order to oversee
- our activities and then we can assess it.
- But I would certainly support a desire
- to make sure that there are enough cleared staff
- to perform oversight, absolutely.

- MS. COLLINS COOK: So transiting to the
- member questions and while this is happening,
- just a reminder there are folks with cards, if
- 4 you have questions that you'd like to submit from
- 5 the public.
- And to keep everyone on their toes, this
- ⁷ time I'm going to start with Pat.
- MS. WALD: Okay, you may be sorry about
- ⁹ that choice.
- MS. COLLINS COOK: I might not be, they
- ¹¹ might be.
- MS. WALD: This is somewhat of a loaded
- question, but it's one that's sort of in the back
- of so much of the work we have done and will
- 15 continue to do.
- You know, I laud all of Becky's
- 17 attempts, and your attempts to inject, Erika,
- 18 your attempts to inject privacy into all of the
- various phases of intelligence.
- But drawing upon what some of the people
- in the first panel said this morning, let me just
- pose a question that, for instance, several of

- the panel members thought collection was a
- ² primary focus of trying to enhance privacy
- interests by limiting collection somewhat, and
- 4 leaving apart any debate about whether or not
- 5 collection by itself can be an injury to privacy,
- ⁶ I guess, and that's collection.
- Also when you get, another expert talked
- 8 about the risks to privacy from aggregating data.
- ⁹ And we found out, for instance, in the 702 report
- we did, when you got to the retention of data the
- analysts might look at it and say, well, I don't
- see any foreign intelligence purpose to this
- 13 piece of data if it came from an innocent person
- who's not the target, but it's conceivable there
- might be one down the line or some other person I
- don't know about, the agency, so therefore, I've
- got to bend to make sure that it's secured.
- So it seems to me one of the basic
- problems here will be, what's the tipping point?
- In other words, assuming good faith on both
- 21 sides, there really is a national security
- interest when you have to make a choice between

- 1 privacy and national security, but the real
- question is, how much and at what point?
- In other words, when we were doing 215
- 4 we were told many times we need a big haystack in
- order to find the needle, and the bigger the
- 6 haystack, the more likely we are to find the
- ⁷ needle.
- But of course a policy judgment has to
- be made at some point. At this point, yes, we're
- going to lose some national security things but
- 11 privacy is more important.
- I guess I want to know what your
- thoughts are about how that decision, which is a
- basic policy decision, but it seems to come up in
- every program that we look at, you know, how is
- it made or how it should be made, even at the
- most general level. You can all take a --
- MR. JOEL: Okay, so I'll start. I'll
- offer some general observations.
- MS. WADE: Yes.
- MR. JOEL: So I think on the collection
- 22 and use and retention point, I would say that

- it's very important to look up each phase of
- that. And that's, in fact, how the intelligence
- 3 community structures its determinations in many
- 4 ways. It's collection, then there's retention,
- 5 and then there's dissimanation.
- And on the collection point --
- MS. WALD: And aggregation.
- MR. JOEL: Right. And then of course
- ⁹ when you aggregate data, you create additional
- 10 risks.
- So there's no question that if your
- concern is to protect privacy, the better way to
- do it, and you're worried about what the
- government's going to do with your data, it's
- always better for the government not to have the
- data. That's the best protection.
- So if the government doesn't have the
- data, there is no risk to privacy from the
- 19 government because they don't actually have it.
- So that's why I think it's appropriate of course
- to focus on collection.
- Once a determination is made that the

- government really needs this data in order to
- carry out an important function, then you're
- 3 shifting to retention. And so there are --
- MS. WALD: Let me just interrupt you.
- ⁵ I'm sorry to do this.
- 6 MR. JOEL: Okay.
- MS. WALD: It's an old habit of mine
- 8 leftover from --
- 9 MR. JOEL: Yes, your Honor.
- MS. WALD: When you say, really needs,
- that's where the rubber hits the road.
- MR. JOEL: Right.
- MS. WALD: Because, sure, it's going to
- be useful. So where the line is between
- something which genuinely will be useful to you
- but will be more of a privacy risk, and the thing
- 17 of, this is really necessary, because. And we
- 18 all know it's going to be drawn differently in
- 19 different case situations.
- But that's what it always seems to sort
- of come down to, and I'm wondering do you have
- 22 any thoughts about how that, which is a policy --

- MR. JOEL: So this is where, and I know
- Becky, but before you used the term tipping
- point, which I think is a very helpful term, and
- sometimes people think of this as a balance or as
- 5 a scale.
- The way that I think of the balance
- ⁷ metaphor as it might apply here is not that
- you're saying, well, that tips it over here so
- ⁹ therefore we're going to do it, that tips it over
- here therefore we're not going to do it.
- 11 Although to some extent, of course, that happens.
- The way that I think of it is that if
- you're going to do something new, a new or
- different collection program, you ask the
- following questions, A, is it lawful? Of course
- it has to be lawful. Is it justified? What is
- the purpose? You know, going to sort of a FIPS
- analysis, what is the purpose for it? Is this
- 19 collection focused on a valid purpose that we
- feel should be pursued and is it important to be
- pursued, whatever the phrasing should be?
- 22 And is your activity tailored to that

- 1 purpose? Are you doing something? Are there
- less intrusive ways of doing it? Is this the
- 3 appropriate way to go about doing it in terms of
- 4 obtaining this information?
- 5 And then what are the risks to that?
- 6 Sort of now going to the other side of the scale.
- ⁷ And how do you guard against those risks? How do
- you mitigate those risks?
- And this is the way that I've always
- thought of it. You know, it actually fits into
- 11 some FIPS kind of models. It also fits into some
- 12 privacy and assessment kind of models.
- But if you look at that overall picture
- it then helps inform you, either the art or
- science side, I don't know, Becky can tell us
- which one that is.
- 17 It helps inform the decision about
- 18 whether this is the right thing to do. And I
- think you have to look at that to tell.
- So if you're just going to do one
- 21 program, well, it's lawful and we think we need
- it. But now you can't figure out, there are

- 1 major risks, but you can't figure out how to
- ² adequatly mitigate those risks, then that will
- 3 tell you one thing about the overall risk of
- 4 doing that activity.
- MS. COLLINS COOK: Alex, if we could.
- 6 MR. JOEL: Oh, I'm sorry.
- MS. COLLINS COOK: That's all right.
- ⁸ And Becky, did you have something specific you
- ⁹ wanted to say in response to this question?
- MS. RICHARDS: Yes, the only thing I
- would say is we've been asking some different
- questions to try and tease out some of this
- conversation as we go through different programs.
- And the questions we've been circling
- around, which are a little bit different than,
- you know, is this lawful. It's more, what is the
- type of the data? How intrusive is the data?
- 18 How broad is the collection? In other words, am
- 19 I obtaining a lot of people who are sort of an
- incidental collection, are not part of the target
- or not? And then what are the stated uses or
- ²² future uses?

- And we've sort of been using those three
- questions to get at, I think, the overall risk,
- which this sort of bubbling it really up is we
- want to stop the government from doing bad things
- 5 to good people.
- And so you know, sort of looking through
- ⁷ those different lenses it helps us do that
- 8 analysis.
- 9 MS. COLLINS COOK: So thank you. David
- 10 -- I'm sorry.
- MS. BROWN LEE: I was just going to say,
- just because you wanted-- all right.
- It's an iterative. I was just going to
- just sort of follow-up on the comments in that I
- think that forcing mechanism of trying to do, of
- having ongoing vetting and ongoing evaluation by
- the right people is where to go, because in
- looking for the meaningful relationships and
- developing those, as opposed to, you know,
- retaining the isolated pieces.
- So I would just say that trying to force
- that mechanism of ongoing vetting is really

- 1 important.
- MR. MEDINE: One of the reasons for
- having the forum today is to get a better
- 4 understanding of what privacy interests are being
- 5 protected by your offices and our agency.
- And Alex and Erika have both been in
- ⁷ either the private sector, in the case of
- 8 Marriott or at least at FTC had a private sector
- 9 focus. How would you compare the privacy
- interests you were trying to protect in your
- prior positions with the interests that you are
- trying to protect now? What are the similarities
- and what are the differences?
- MR. JOEL: So I actually think there are
- a lot of similarities, but there are of course
- some important differences as well.
- So on the similarity side, and I think
- 18 privacy officers and people in all kinds of
- organizations, be they private sector or other
- government agencies, share a similar challenge or
- 21 problem set, which is when your organization
- wants to do something either for a business

- purpose or for an authorized statutary purpose,
- and in order to do that you need information.
- And for businesses this is typically
- 4 information about customers or potential
- 5 customers. And then you want to do something
- 6 with that information to carry out your lawful
- ⁷ activity. So it's a given that your organization
- 8 will be obtaining and using personal information
- 9 in many cases.
- And so then the privacy officers'
- 11 challenge is then making sure that that activity
- is conducted in a way that maintains your key
- 13 trust relationships.
- There are different ways of framing it,
- but I think that's generally speaking what
- happens.
- And so for a business perspective, what
- 18 you want to make sure you're doing is delivering
- value to your customer and that you're not using
- that information for inappropriate means or ways
- that are going to essentially get your customer
- upset and have your customer take his business

- 1 elsewhere.
- And so a lot of those things are
- 3 similar. I think that the key distinction for a
- 4 business is of course that it has the ability to
- 5 disclose a lot about what it's doing in terms of
- 6 obtaining that information. And the value that
- ⁷ it's providing is also something that gets
- immediately, it should be immediately apparent to
- ⁹ the customer.
- To the extent that the value is further
- down the chain and the customer doesn't see it
- that much, but is aware that the information is
- being collected, that impacts the trust that the
- customer has with the business.
- I think from an intelligence community
- perspective, it's hard for us to demonstrate the
- value. What are we doing with the information?
- And so as a result when people are worried about
- information being obtained by the intelligence
- community, the value to them seems inchoate, yet
- the risks seem very real. Like, well, my freedom
- 22 could be impacted if the government misuses this

- ¹ information.
- We can reassure people we have methods
- in place to make sure that the information will
- not be misused, but I think, and we need to do a
- better job of that, but I think the other side of
- that equation is we have to show, better show
- what we're doing with the information.
- 8 And of course for intelligence agencies
- 9 some of the most tightly held secrets are the
- successful use of intelligence, because we don't
- want our adversaries to know that that method was
- 12 successful.
- MS. BROWN LEE: Okay. So just to
- quickly answer your question, so I was also in
- the private sector at a law firm and practicing
- 16 privacy.
- Here's where they're similar, whether
- it's clients, or even from a government
- prospective, people tend to be reactive to
- ²⁰ privacy.
- 21 And one of the things that I find the
- biggest challenge is to be proactive. And it

- 1 means sometimes taking unpopular positions,
- whether it's with clients or internally within my
- organization. But sort of having principled
- 4 reasons for doing that, and if not forcing
- ⁵ putting, you know, very strong arguments to do
- 6 what you think is the right thing, I think is
- where it's simialar and where it's hard but
- 8 interesting.
- 9 MR. MEDINE: Becky, you talked about
- catagorizing certain types of information as
- 11 being sensitive.
- 12 In our morning discussion there was a
- 13 lot of talk about the mosiac theory where there
- may be individual bits of information that are
- innocous on their face but in combination they
- present a perhaps sensitive profile of someone's
- activities, thoughts and so forth.
- Do you lose something if you focus on
- what seems to be sensitive information and not
- take into account the potential combinations of
- 21 information?
- MS. RICHARDS: So actually the goal is

- to take in to all those combinations. So the
- idea and where we've been looking at is that it's
- yery difficult. You know, we want to push folks
- 4 and I will say that this is an uncomfortable
- ⁵ place to be as a privacy person. This is sort of
- 6 where I'm like, well, it'll depend.
- But if we look at where big data is
- 8 today, there is a lot of data and it's very
- ⁹ volumonous and it's a lot of discrepancies. And
- if we can start to define, which is sort of what
- 11 I felt like we heard in the second panel, even
- if -- and this is where sort of I think we're
- 13 going to try and push NSA, is if we can start to
- define and put some mathematics behind it.
- So that, for example, if you have
- vaguely anonymous or slightly de-anomoized data
- over here and over here and the computers start
- to put them together, we would want the system to
- then pop something to say, hey, look at this
- before you decide to go forward.
- So the idea is technology is supporting
- the privacy analysis by looking at whether or not

- ¹ the math underneath it can work.
- And so you're going to have to make some
- ³ really hard choices. Do I think health data is
- 4 more risky to privacy than my address? And
- ⁵ everybody gives the example of, well, then you
- 6 have the violence against women or, you know,
- ⁷ something along those lines.
- But at some level if we deal with only
- those edge cases, we're not going to move
- forward. And I think the value, we will be
- losing some of the value, both from a privacy
- perspective, as well as from a technical
- perspective.
- Because we're sort of in this art form
- of looking at each individual case, which I
- recognize at NSA, I'm not going to be able to
- 17 look at every single, little thing. We want a
- system to be able to identify the things that
- 19 need additional analysis, that need that
- ²⁰ additional judgment.
- But what I don't want to have happen is
- have us backed into a place where the system is

- doing things that we would find unacceptable
- because we didn't sort of build something in to
- 3 help with that.
- MS. COLLINS COOK: Thank you. Rachel.
- MS. BRAND: Thank you all for being
- 6 here. For those of you who have been here all
- day you'll know that this is a little bit of a
- 8 hobby horse of mine. But I want to ask about the
- ⁹ FIPS and why you are purporting to apply them,
- although you can't really apply them.
- So I gather, and Ms. Richards, this is
- directed to you, at least initially. And I
- commend you for publishing the paper on targeted
- collection under 12333. And you said that you
- were applying the FIPS, and I gather you were
- talking about the 2008 DHS iteration of the FIPS.
- But then you said that, for example, the
- individual participation FIPPs can't really apply
- 19 to your activities, and the transparency one can
- apply in a very limited way.
- I guess I'm wondering whether it doesn't
- make sense to come up with a new set of

- principles that applies to survellance activities
- of the government? Because if you look at the
- 3 DHS FIPS, the transparency one as articulated in
- 4 this document really can't apply to you because
- 5 it's talking about providing notice to the
- 6 individual regarding collection. That's
- obviously not going to take place. Individual
- 9 participation really can't apply at all.
- 9 MS. RICHARDS: Correct.
- MS. BRAND: Some of these other ones are
- 11 very, very important. Purpose specification is
- very important. Miniminization, data security,
- some of these are important.
- But yet, this doesn't at all address
- things like thresholds, evidentary thresholds for
- collection, which are required obviously by law.
- 17 But if you're talking about principles that are
- supposed to sit on top of the fundamental legal
- 19 requirements, you should talk about thresholds.
- 20 And there are some other principles that don't
- 21 come into play here.
- So I'd be interested in knowing why you

- decided to apply the FIPS and if you've given
- 2 some thought to coming up with some new
- ³ principles.
- I don't mean to critize this for DHS's
- ⁵ purposes because DHS has a lot of functions that
- involve voluntary interaction by an individual
- with the government where this makes a lot of
- 8 sense. So but you're in a different positoin
- ⁹ than that, obviously.
- MS. RICHARDS: So I guess what I would
- say is it's a beginning place, and I've sort of
- 12 stated that a couple of different times because I
- wanted to start with something.
- And so from my perspective, I guess I
- want to take the parts of the FIPS that work
- well, which would be basically the bottom six of
- 17 the DHS ones and then look at how we can work
- 18 those through.
- So what I would say is sometimes there's
- 20 analysis that needs to be done at an enterprize
- level. So it's useful for me walking into the
- 22 agency, which may be readily apparent to

- everybody, but it was just useful to go through
- the process and say, hey, here is sort of one
- framework that we think about for privacy, and as
- an enterprise we don't do the first two.
- One of the questions that led me to ask
- in some of the conversations I've had with
- academics and advocates is to say, okay, we don't
- 8 do transparency in the traditional sense and we
- 9 don't do individual participation. Is there some
- 10 proxy? Is there some additional thing that we
- should be doing, given that?
- 12 And I think that gets to your question
- 13 of, well, are there other things that should be
- underpinning these? And that's where we're
- starting to work through those questions.
- So I think it was very beneficial to
- start with that as the beginning one and then use
- the remaining six principles as the basis for
- 19 some of these questions.
- Part of the problem though I will tell
- you with the FIPS is they don't give you a
- judgment. They don't tell you, well, this is

- good enough or that's bad enough, which sort of
- gets to your evidentiary purpose. And that's the
- 3 place where we are trying to then look at the
- 4 data. What are the risks to the data?
- We spend a lot of time now talking
- 6 about, well, what is the exact risk to this
- ⁷ program to privacy and civil liberties? And so
- 8 we're still working through those and having a
- 9 lot of really fun and intellectually stimulating
- 10 conversations about what are the right questions
- and how do we do that for an intelligence agency
- 12 at NSA.
- But I would just say that it, for us,
- was a beginning place. I don't think that's it's
- 15 necessarily the ending point, but it was
- someplace to start with. And I don't want to
- 17 sort of throw everything out and start with, I
- don't know. You know, you have to start
- 19 somewhere.
- MS. BRAND: Okay. Do the other
- 21 panelists want to say anything about that? Alex?
- MR. JOEL: I would just say that even

- though the first two do not directly apply,
- ² certainly not as written by DHS, they provide
- ³ useful measures for us to determine to what
- 4 extent does this raise privacy issues and in what
- ⁵ areas.
- So that is, I think it's very helpful to
- ⁷ use as a guide in the way that Becky has been
- ⁸ using it at NSA.
- I like the idea of developing a
- statement of principles that would apply to the
- intelligence community. So I'll take that path.
- MS. BRAND: I think I probably don't
- 13 have time for another question, but I would
- suggest if you're going to engage in that
- exercise that you look at the threshold question
- and that you also look at oversight because
- these, you know, they talk about accountability
- and auditing, but creating a paper trail is not
- the same thing as effective oversight. And
- obviously, as I said in the previous panel,
- oversight is extremely important in this context.
- 22 So just food for thought.

- MS. RICHARDS: And I think it's just
- important that you don't have a check box. I
- mean part of the problem I think with the FIPS
- 4 also is it leaves itself to a little bit of a
- 5 check box process.
- Do I have a privacy statement? Yep, I
- ⁷ got a privacy statement. Okay, am I doing
- 8 everything in there? Yep, okay, I can do that.
- 9 As opposed to these sort of questions
- of, should I be doing that? And that's I think
- where having an individual at the agency whose
- focal point is this, really benefits the agency
- in terms of that conversation because it can very
- quickly devolve to, I checked it off, I'm good.
- You have no privacy, but I'm good.
- MS. BROWN LEE: And I would just say
- that the oversight perspective has to also be
- iterative and changing because I think as
- technology allows us to, you know, collect more
- data and in different ways and different data
- points that the oversight of it has less meaning
- if you're not also adapting on that side as fast

- as we are adapting to the technological changes.
- MS. COLLINS COOK: Thank you.
- Jim, do you have some questions?
- 4 MR. DEMPSEY: Thank you. Thank you to
- 5 the members of the panel.
- I have some questions that I want to
- ⁷ ask, but I saw there are a lot of audience
- guestions. Was there one or two that stood out
- 9 particularily? Technically, we only have five
- more minutes to go on this panel, so I'm happy,
- 11 Beth, to have you ask one or two of the audience
- 12 questions.
- MS. COLLINS COOK: Sure, and I think you
- should know you have won the jackpot thus far on
- ¹⁵ audience questions.
- Alex, this one goes to you and it draws
- on a remark from a previous panel.
- Why can't the IC inform the American
- people about how many phone records were
- 20 collected pursuant to Section 215 and make
- similar public disclosures regarding the breadth
- of U.S. person collection under 702 and E0 12333?

- So the executive order, understanding
- that you're not targeting necessarily U.S.
- persons, but the U.S. person incidental
- 4 collection.
- MR. JOEL: So that's a good question. I
- 6 don't want to duck it. I'll just say that I am
- ⁷ going to in certain ways.
- No, but I don't want to, I guess I'm not
- ⁹ going to get into the specifics of like 215 or
- ¹⁰ 702, etcetera.
- What I'll say is that there are two
- 12 challenges. I understand the interest and I
- understand the importance. One is technical
- capability. Can you, in fact, count it? And for
- some things, some activities, you should be able
- to count. But for some other ones, they
- inherently involve challenges.
- I know that one of the PCLOB
- recommendations in the 702 report was, in fact,
- to count some of the 702 collection that involves
- U.S. persons. So there are some inherent
- 22 challenges in doing that.

- 1 From a national security perspective
- what I'll say is what I have heard internally as
- we have pursued these kinds of questions is that
- 4 providing that kind of information can, in fact,
- ⁵ put at risk some kinds of collection, especially
- 6 if you track it over time.
- An adversary, a sophicated adversary can
- 8 put the information together in terms of the
- ⁹ volume of collection in one particular area and
- then draw some conclusions about what
- specifically is being obtained, what are the
- 12 specific channels that are being watched, and
- therefore change behavior.
- So our job from a transparency
- perspective is to continue to discuss that
- internally and see, well, you know, are there
- ways of mitigating that? What can we, in fact,
- disclose in this area? Because it's of strong
- 19 interest.
- MS. COLLINS COOK: So, Erika, I'll
- direct this next one to you because you mentioned
- that part of the civil liberties protections and

- privacy protections are consequences for
- ² wrongdoing.
- 3 So the question from the audience is, in
- 4 the case of a privacy violation sufficient
- ⁵ remedial measures are critical. What, if
- anything, do you think needs to be done, either
- ⁷ statutorily or administratively to strengthen
- 8 existing remedial schemes?
- 9 MS. BROWN LEE: So, yeah, I do think
- that the remedies for Privacy Act violations or
- 11 for privacy violations are, you know, as I said
- in my remarks everything could be examined and
- 13 looked at for approval.
- I was focusing my remarks on FBI. So of
- 15 course they have their own investigative unit
- that reviews. So if there's any particular
- activity that an agent engages in, for example,
- that is, you know, collecting information in
- violation or specifically because of First
- Admendment purposes, that's subject to review and
- 21 discplinary action.
- With regard to individuals, I agree. I

- mean we talked about how the FIPS doesn't really
- have as meaningful of really a guide for law
- 3 enforcement either.
- I think, you know, it's not something
- that I can do, but certainly it's been attempted
- before to remedy the Privacy Act or to amend it.
- We are, the administration is committed
- 8 to looking to expand the protections of judicial
- 9 redress for non-U.S. persons, and DHS has a
- 10 policy of doing so administratively.
- But I think statutorily it's a hurdle.
- 12 I think it's something that I would be willing to
- have a conversation to further that.
- MS. COLLINS COOK: So just to keep this
- even across the board, Becky, this one is for
- ¹⁶ you.
- And I think implicit in this question is
- 18 a very interesting premise. Do you anticipate
- that wide swaths of data will no longer be
- 20 collected now that you are asking questions about
- whether they are really needed and the civil
- 22 liberties downsides?

- So I would say the premise is that it's
- your job to shut it down, which I think it's a
- ³ widely shared premise.
- 4 And I think the basic question is, do
- you think you're going to be effective?
- 6 MS. RICHARDS: So I think that also
- ⁷ starts with the premise that the collection we're
- 8 doing currently -- that's starting with the
- 9 premise that we're collecting too much
- information today.
- And I think what I would say is that
- what we're working on is sort of a premise, so if
- NSA is filled with a lot of people who do math
- 14 for a living, we're in the process of third grade
- math, which is folks need to show their work. So
- they need to show why they're doing what they're
- doing so that then we can have those
- 18 conversations.
- I don't want to presuppose we're going
- to do more, or less, or either way of those. But
- I do think that what we haven't done well is
- explain what we're doing.

- And if you sort of consider that NSA has
- a long history of saying absolutely nothing to
- anyone, and in the last year and a half we've had
- 4 to create a voice for ourselves to explain what
- it is we do, and recognize that most people,
- there are a lot of Ph.D.s in math at NSA who
- don't necessarily take well to speaking in
- 9 public. It's a work in progress.
- And so my hope here is not to be judged
- by how much we turn on or turn off, but by
- demonstrating what the value is to the country in
- terms of what we're doing and demonstrating that
- we're protecting civil liberties and privacy.
- MS. COLLINS COOK: So thank you all for
- your remarks and your active back and forth on
- the questions.
- MR. MEDINE: And we'll be taking a 10 or
- 18 15 minute break, and at 2:45 we'll resume with
- the private sector's views on these issues.
- ²⁰ Thanks.
- 21 (Off the record.)
- MR. MEDINE: Good afternoon. Thanks for

- everyone's endurance who's been here all or most
- 2 of the day.
- This is our final panel today, but an
- 4 important panel on what the private sector has
- be learned about privacy and how that might relate
- to the considerations we go into with regard to
- ⁷ national security issues, and Rachel Brand will
- 8 moderating.
- 9 MS. BRAND: All right, thank you, David,
- and thank you to all our panelists for being
- 11 here.
- The way we've structured the day is that
- the first panel this morning had to do with the
- theoretical underpinnings of privacy and
- exploring what interests underlie privacy. The
- second panel had to do with technology. The
- third panel was a government panel.
- And this last panel is supposed to be
- 19 focused on solutions, and particularly those
- solutions that folks in the private sector might
- 21 be able to suggest.
- So what we'll do here logistically is

- each panelist will start with up to seven minutes
- of remarks. And for the panelists' benefit, Sam
- 3 Kaplan is sitting in the front row here with
- 4 yellow and red cards. So when he holds up the
- yellow card, you'll know you have two minutes
- 6 left, so please pay attention, and the red card
- ⁷ means that your time is up.
- At that point, as the moderator I will
- 9 ask about 20 minutes of questions, and then each
- of my fellow Board members will have five minutes
- of questions, and then we'll open it up to
- 12 questions from the audience.
- And as with the previous panels, when I
- start to ask questions some of our staff members
- will stand up in the back, and Lynn Parker Dupree
- in particular, and Prem, will stand up and hold
- up cards and you can go get yourself a card,
- write down your question and then the staff will
- 19 pass it up here.
- So we'll just go down the row and we
- will start with Professor Cate. I am not going
- to go into length on their biographies because I

- think they're all available to you.
- But Professor Cate is a Professor at the
- University of Indiana School of Law, and he's
- 4 been on a number of previous boards and
- ⁵ commissions on privacy.
- And so, Professor Cate, let's start with
- you.
- MR. CATE: Thank you very much. This is
- 9 the time I think to say that I'm colorblind so
- 10 I'll have no idea what cards you're holding up.
- 11 So perhaps you'll wave them in a definitive way
- and I will pay attention.
- So first of all, I was sorry not to be
- here for this morning, but the last panel was
- absolutely superb, and it's a privilege, both to
- be here, and I really want to applaud the Board
- for taking up this I think really difficult, but
- fundamental, issue about what is privacy and how
- in practice might we go about protecting it, both
- in the private and the public sectors.
- I want to really just offer some
- observations, as opposed to any specific, if you

- will, recommendations or conclusions.
- One, and this was touched on in the last
- panel, I think the FIPPs are frankly not
- tremendously useful.
- I'm not suggesting abandoning them,
- 6 which is a big change for me. Ten years ago I
- wrote a book chapter called, the death of the
- 8 FIPPs. But fortunately, I've gained a little bit
- ⁹ of knowledge here.
- But I think we use them almost
- talismanically, like we can roll out these eight
- principles, or depending on what list of FIPPs
- you use, and that that will get us somewhere.
- 14 And that far too frequently, both in the private,
- and certainly in the public sector, they really
- don't get us anywhere.
- What we end up is we end up, just like
- talked about in the last panel, looking for
- substitutes for the FIPPs. Well, we can't have
- consent, what could we have, rather than asking
- what was the purpose to be served in the first
- ²² place?

- And maybe consent's no longer relevant
- as a tool to achieve that purpose, rather what
- ³ are we trying to actually do here? Really the
- 4 question you've been asking all day, what are we
- trying to protect? What do we think protecting
- 6 privacy really means?
- I say this, by the way, about the FIPPs
- in part because I'm not sure that they've ever
- ⁹ worked terribly well, and certainly in the U.S.
- environment where they've largely come to mean
- 11 notice and choice.
- 12 I'm not sure that they work well in a
- world of massive data, whether we call it big
- data, or whether we call it just high volume
- data, but the notion of a sort of FIPPs-like
- approach, particularly with a focus on the
- individual when the broader issues may be,
- 18 frankly, societal. They may be the impact on the
- economy. They may be the impact on civil
- liberties, not of one person but of everybody.
- 21 And I don't know that the FIPPs help focus us in
- 22 a useful way on that.

- And then frankly, I think the FIPPs have
- led to some sort of silly results. And you know,
- 3 I would just mention I've always been surprised,
- for example, the Department of Homeland Security
- ⁵ privacy impact assessment on border searches of
- 6 electronic devices, which focused a lot on notice
- ⁷ as a privacy protection.
- 8 Well, at the point that your device has
- been seized from you and its contents copied it's
- difficult to think that notice is meaningful
- protection. It may be necessary, but whether
- it's protection or not, I think it's not.
- Second point, one of the things we are
- seeing emerging in the debate in the private
- sector, and we see this especially in Europe in
- the context of discussing the general data
- 17 protection regulation there, is greater focus on
- 18 risk management, or risk assessment and risk
- management.
- And I don't mean to use this just
- because it's sort of the jargon of the day but
- ²² rather because risk management is an incredibly

- valuable tool that in privacy we are unbelievably
- far behind on. You know, security we have much
- 3 clearer ideas what risk management mean. Privacy
- we really lack that understanding.
- And part of the reason is because we
- don't know what risks we're guarding against.
- We're very unclear what are the harms, what are
- 8 the impacts, what are the negative effects that
- ⁹ we think we are balancing, if you will, with the
- positive outcomes of the use of data or what have
- ¹¹ you.
- And so one reason I think the risk
- management approach offers a lot of value in both
- 14 public and private sector is that it makes us
- stop and say, what is it we're trying to
- 16 accomplish? What are the positive benefits and
- what are the potential negative impacts, not
- measured in terms of FIPPs, but measured in terms
- of actual impact on individuals, or on society,
- or on the economy, as we think about it.
- When using risk management, or if you
- hate risk management, in either case, third

- point, I think there's a lot of reason to focus
- ² more attention on use of data.
- And this has been a real weakness of the
- ⁴ U.S. legal system. Those of you who have
- ⁵ suffered through law school know that the Fourth
- 6 Amendment has almost nothing to say about use of
- ⁷ data whatsoever.
- In fact, you can have illegally seized
- 9 data that the court acknowledges is illegally
- seized and they will still allow it to be used
- someplace else because there would be no
- disincentive for the collection, it's only the
- collection in the Fourth Amendment that Supreme
- 14 Court jurisprudence has been focused on.
- And for this reason I think we really
- would be better to be thinking more about
- reasonable and effective limits on use. And in
- 18 fact, I think that's what the public most
- 19 commonly cares about.
- And one of the practical reasons for
- that is because there's almost always a
- legitimate reason to collect the data. There's

- always some reason, there's some employment
- reason, there's some security reason, there's
- some private sector reason. You know, Verizon
- 4 had a reason to collect this data. And then the
- ⁵ question was, who could access it and how could
- 6 it be used?
- But our legal system has focused
- 8 enormous attention on collection, and then once
- the data are in the government's storehouse then
- we feel that the data are more commonly out of
- 11 control. And I think that is a critical area to
- 12 focus on as well.
- Fourth, as I've mentioned, I think the
- 14 Fourth Amendment, while it's a critical legal
- limit and I certainly concur -- that's yellow,
- 16 right? Yes? Thank you. And for the rest of
- you, you'll know I just got a yellow card.
- I think the Fourth Amendment of course
- is a critical legal limit and we must of course
- observe it. It's not a very useful guide for
- telling you what to do in the future for a
- positive analysis of privacy issues.

1 And I think we should again be careful 2 Too often in our rhetoric we say about that. 3 well, it's permitted under the Fourth Amendment as if that tells us anything, other than it is 5 not illegal under the Fourth Amendment. But it 6 doesn't tell us anything about either the ethics, or the desirability or what have you of doing it. And then fifth, I would just say in 9 almost all of these areas, and I understand in 10 national security this seems particularly maybe 11 odd, I think redress is something we need to 12 continue to focus on. 13 We see many uses of data in the 14 government setting and in private sector which 15 are done without regard to redress, with just 16 sort of, well, if it affects a person 17 inaccurately every now and then, what does it 18 really matter? We'll deny boarding to people on 19 airplanes, or we'll provide extra security for 20 the wrong people. 21 This is not an efficient use of 22 government resources and it's not a good way to

- ¹ think about privacy.
- 2 And I think we should be very clear in
- those rare exceptions where we say there might be
- 4 no redress available here for the individual, in
- which case we now have to provide it through
- other means, inspector generals, or the PCLOB, or
- other ways of approaching it.
- But at all times we should be thinking
- 9 about redress, not just because of the rights of
- the individual, but because of the interests in
- ensuring that the system works as advertised and
- 12 as it should. Thank you very much.
- MS. BRAND: Thank you very much. Our
- 14 next panelist is Harley Geiger. Harley Geiger is
- 15 Advocacy Director and Senior Counsel at the
- 16 Center for Democracy and Technology, and he
- focuses on issues related to civil liberties and
- government surveillance, computer crime and
- 19 cybersecurity. Thank you for being here.
- MR. GEIGER: Members of the Privacy and
- 21 Civil Liberties Oversight Board, thank very much
- for inviting me to speak at your meeting today,

- and thank you also for your excellent work on
- ensuring protection for privacy and civil
- 3 liberties in national security and terrorism
- 4 programs. And congratulations on having one of
- 5 the best acronyms in town.
- When it comes to evaluating privacy
- ⁷ protection, the Center for Democracy and
- 8 Technology believes that the Fair Information
- 9 Practice Principles are a very important
- framework for both government and the private
- 11 sector.
- Now you can add other privacy frameworks
- on top of that. We certainly do not disagree
- with Professor Cate that societal impact is a
- very useful consideration, and we certainly agree
- that protections focused on the purpose of data
- 17 collection are also useful.
- But we view the FIPPs as an
- 19 indispensable framework for evaluating privacy
- protection with data collection practices.
- Now the individual principles of the
- FIPPs, as you know, are overlapping and mutually

- dependent on one another. It is a framework.
- 2 It's not a smorgasbord that you can just choose
- and pick, at least not unless you don't want
- 4 robust privacy protection.
- 5 And there is obviously some discussion
- in the private sector about doing away with data
- 7 collection limitations or the data minimization
- ⁸ principle of the FIPPs, seeing as how we are now
- ⁹ all in an age of big data.
- But in the time that you've given me, I
- want to address this head-on in the context of
- 12 government surveillance.
- First, CDT believes that there still
- should be collection limitations on private
- sector data collection, and that the data
- minimization principle of the FIPPs should apply
- to the private sector.
- Second, the government should not take
- its cues entirely from the private sector when it
- 20 comes to national security surveillance. Data
- 21 collection from the private sector is
- ²² fundamentally different from national security

- ¹ surveillance.
- Therefore, even if the private sector
- were to collect data in a relatively unrestrained
- 4 manner in some alternate universe, then
- ⁵ intelligence agencies should still nonetheless
- 6 not follow suit.
- 7 The missions of the private sector and
- 8 the national security functions of governments
- ⁹ are totally different. That should go without
- 10 saying. The private sector typically does not
- use the data that it collects to detain or take
- 12 kinetic action against the individuals as part of
- 13 its mission.
- Several major private sector companies
- have repeatedly responded to public outcry over
- privacy with enhanced transparency and privacy
- 17 controls.
- The national security arms of government
- are not as transparent or responsive and are not
- likely to be.
- Many major companies, in addition, allow
- or are required by law to allow consumers to

- limit the collection of information about them.
- ² More and more services are differentiating
- 3 themselves on the basis of strong privacy
- ⁴ protection.
- 5 And of course, individuals can choose
- on not to participate in a commercial service as a
- means of limiting direct data collection about
- 8 them. But data collection for national security
- 9 purposes does not permit any meaningful choice.
- So this is not to laud private sector
- data collection practices because CDT does view
- them as generally insufficiently protective of
- 13 privacy.
- But because of the differences that I
- just briefly listed, and other reasons, even if
- the private sector fails to robustly apply the
- 17 FIPPs government agencies should not follow suit.
- 18 If anything, because of these
- differences government should strive for a more
- strict and consistent application of the FIPPs
- than that of private sector data collection.
- 22 And so I have a small set of broad

- recommendations to make.
- First, the government should place
- 3 greater emphasis on applying the data
- 4 minimization principle of the FIPPs. Back-end
- ⁵ minimization procedures alone are not sufficient.
- ⁶ Front-end minimization is also critical.
- 7 Trust is breached at the point of
- 8 collection. Once the government collects
- ⁹ information, non-statutory internal restraints on
- 10 access and use can fall away like sand castles on
- a beach. We saw this happen with the 702
- backdoor search loophole.
- So surveillance should be restricted at
- the front-end by narrowing limiting the
- collection of data to what is directly needed to
- accomplish a specific purpose.
- The data should then be retained only as
- long as is necessary to fulfill that purpose, and
- 19 the data should be destroyed unless a
- determination is made that the data are needed to
- 21 accomplish the specific purpose.
- The specified purpose of data collection

- itself should be subject to meaningful
- ² restriction.
- For example, limiting the scope of what
- 4 is relevant under Section 215, or the definition
- of foreign intelligence in Executive Order 12333.
- So the goal should be overall to move
- ⁷ from mass data collection to targeted data
- 8 collection of both U.S. and non-U.S. persons.
- 9 Second, the government should provide
- much greater transparency regarding the
- interpretation of surveillance laws. Section 215
- of the PATRIOT Act exemplifies this.
- Nobody was surprised that the NSA is
- collecting phone records. What was surprising
- was that the NSA has secretly interpreted Section
- 16 215 to allow for the collection of all phone
- 17 records in the entire country.
- This is bad data minimization. And yet,
- a fair reading of the statute does not seem to
- grant them with this authority.
- So declassification of FISA court
- orders, or, when necessary, summaries of opinions

- would substantially boost transparency. We
- should not be a nation of secret laws.
- Third, the government should provide
- ⁴ greater transparency around the extent and scope
- of requests for data under national security
- 6 authorities. This includes government reporting
- ⁷ about its national security surveillance
- 8 activities, such as how many requests were made,
- ⁹ under which surveillance authorities, and for
- what type of data, as well as how many U.S. and
- non-U.S. persons were affected.
- The government should authorize the
- 13 private sector to make similar reports.
- 14 Information is power and privacy is
- control of information. An entity possessing
- information about an individual has power over
- ¹⁷ that individual.
- Large scale government collection of
- information about individuals threatens the
- relationship between citizens and the state
- because it upsets the balance of power that
- supposedly exists in democratic society.

- 1 Therefore, CDT urges PCLOB to recommend
- that the government recommit to a robust
- 3 application of the Fair Information Practice
- 4 Principles, as well as other considerations,
- ⁵ regardless of what the private sector does, with
- 6 much more targeted data collection and greater
- ⁷ transparency. Thank you.
- MS. BRAND: Thank you. Our next
- 9 panelist is John Grant. Mr. Grant is a Civil
- 10 Liberties Engineer at Palantir Technologies, and
- 11 he previously served on the staff of the Senate
- 12 Homeland Security Committee where, among other
- things, he oversaw the Department of Homeland
- 14 Security. Thanks for being here.
- MR. GRANT: Thank you very much, and
- thank you for the invitation to speak today.
- As I never tire of telling people, I was
- 18 a congressional staffer who worked on the
- 19 legislation creating PCLOB 2.0, so I take a pride
- of parentage in the Board and I'm sure it's every
- 21 parent's dream to one day testify in front of
- their children.

1 I know that the Board and a lot of 2 people are familiar with Palantir so I'll spare 3 everybody the extended commercial. Just suffice it to say, Palantir builds a data management and 5 data analytics platform that works with data. 6 We started in the law enforcement, 7 intelligence space and have expanded to 8 deployments around the world and in a variety of 9 contexts in the financial sector, medicine and 10 elsewhere. 11 A core tenet of Palantir is that our 12 technology isn't successful if in the course of 13 achieving an organization's analytic mission 14 we're not also able to be deployed in a way that 15 protects privacy. 16 And that's something that the founders 17 of the company instilled from day one, and that 18 is why my job exists, a civil liberties engineer. 19 Well, one of the things I learned when I 20 went to Palantir, and this is different from the 21 Hill certainly, is when you walk into a room and 22 you say to engineers, I'm worried about this

- thing you're building, it creates a privacy
- problem. The response is, oh, okay, how do I fix
- 3 it? Which is not often what you get sometimes
- 4 when you raise these things in other places.
- 5 So it's our job as a civil liberties
- 6 engineering team to come up with suggestions for
- ⁷ how to fix it.
- I am a lawyer, as you may have guessed,
- 9 so I do not necessarily possess a lot of
- technical skill. So the main role for us is to
- translate between the lawyers and the engineers
- 12 and back.
- So what I wanted to focus on today a
- little bit is some of the technology at a high
- level, and then I had some actually suggestions
- 16 for moving forward that I think are actually
- ¹⁷ fairly low hanging fruit.
- So just briefly to provide a little
- context, as I said, Palantir is data management
- and data analytics, so we're not dealing with the
- 21 collection of data. So this gets more to
- 22 Professor Cate's point about the use of data.

- And we have two sort of high level
- ² categories of technology that deal with managing
- or protecting privacy with the use of data, and
- 4 that's access controls and oversight mechanisms.
- But I want to start by pointing out, and
- I think this is something to keep in mind, that
- ⁷ just as technology has expanded the power of
- 8 surveillance today and the amount of data that
- 9 can be collected, it's also significantly
- expanded the level of privacy protection that is
- 11 available at the agencies.
- 12 If you imagine 50 years ago if there was
- 13 an FBI file, this was probably pieces of paper in
- 14 a Redweld sitting on a desk somewhere, or maybe
- locked in a desk drawer, hopefully locked, or
- maybe in a dusty basement archive or something
- 17 like that.
- And you know, there'd probably be very
- 19 limited tracking of where the file was, you know,
- hopefully a log book with a name and who had the
- 21 record. But who knows?
- 22 And anyone who accessed the file would

- be able to see whatever was in the Redweld.
- You'd just be able to rifle through it and you
- 3 could see anything, even if it wasn't directly
- 4 relevant to what you needed.
- Oversight into how the file would be
- 6 used would really be nonexistent. You wouldn't
- ⁷ see exactly who added information to the file,
- 8 who deleted information from the file. And
- ⁹ deletion would largely be hopefully a burn bag or
- a shredder, probably just crumpling it up and
- throwing it in the trash. A more precise
- deletion would be a black magic-marker redacting
- ¹³ a few points of information.
- Today. So today technology allows us to
- do a lot more management of data and oversight,
- and management at a granular level, and that's
- what the access control point, which is you can
- now build access controls to manage data very
- 19 precisely on a data point by data point basis.
- And you can do it in a more nuanced way.
- You don't have to make a choice between access or
- not access. There are ways to sort of have

- 1 gradations of access. You can make the access
- ² controls dynamic.
- 3 So there's a lot of options. And the
- 4 many options you have to configure those access
- ⁵ controls give you a near infinite variety of
- options in how to manage the data, who can see
- ⁷ the data and what they can do with the data.
- The other point is oversight mechanisms,
- 9 and this is really thinking a lot about audit
- 10 logging and also using technological, electronic
- work flows to control exactly how data flows
- 12 around an organization, and who can see data, and
- exactly what kinds of analysis they can do with
- it, even automating, or at least hardwiring in an
- approval chain for use of data and things like
- 16 that.
- And these can be very detailed. So the
- 18 hardwired approval process and things like that,
- that can be very complex. It can involve
- 20 multiple actors, it can involve multiple
- 21 stakeholders.
- 22 And then the auditing of how data is

- used itself can be incredibly granular and
- ² incredibly detailed.
- And I'm skipping over a lot here, but I
- 4 want to get to some of the other points. But
- ⁵ just these two capabilities are a significant
- 6 improvement of what existed before and can get us
- 7 a long way.
- And there are things that exist today.
- 9 Now, I'm obligated to say that of course Palantir
- does these the best, but these are not
- technologies that are exclusive to Palantir. And
- they can be deployed and they can be used in a
- 13 lot of different contexts.
- So what is the problem today, and why
- aren't these capabilities being used more than
- they could be and at the level that we think they
- 17 could be?
- A couple of things. One issue is
- 19 technical awareness. Lawyers don't know
- technology and engineers don't know law, and you
- need people who know both of these things to be
- 22 able to make the decisions of how to use these

- technologies, how to incorporate them effectively
- ² into programs.
- Lack of resources. You need people who
- 4 can actually manage the data. And we talked
- 5 about this in the earlier panel. Alex Joel has a
- 6 very small staff. Erika has a very small staff.
- And they're managing huge amounts of data and
- huge organizations. They need resources, they
- 9 need infrastructure to actually be able to do
- 10 this.
- Privacy is hard. How exactly do you
- look at an audit log? How do you use it
- 13 effectively? How exactly do you manage access
- controls at this data point by data point level,
- especially when you're dealing with mass amounts
- of data?
- And the last one is death by anecdote.
- 18 The argument, the debate, the cost benefit
- analysis right now tends to be the national
- security sector saying one time we caught this
- 21 bad guy using this information, and the civil
- liberties community saying one time this unjust

- thing happened to a person because of this
- ² program.
- There needs to actually be a much more
- 4 -- you can't just make this argument on anecdotal
- ⁵ grounds. You have to actually look at the data
- 6 and you can find out more specifically how these
- ⁷ programs are working, how effective they are.
- 8 So solutions. Obviously we suggest some
- 9 of the solutions in listing the problems.
- 10 Education. I think, and Palantir
- 11 actually sponsors fellowships with Paul Ohm's
- 12 Silicon Flatirons Project and other places to
- make sure lawyers can learn technology and that
- engineers can learn law.
- Engineers don't have to be lawyers, but
- it actually should be a requirement to have an
- engineering ethics program and to have courses
- that teach engineers privacy, because they're
- going to build the technology that's going to hit
- the streets and it's going to be months or years
- 21 before the law catches up.
- So shouldn't engineers be able to catch,

- 1 you know, how what they're building is going to
- ² affect privacy and be able to start thinking
- 3 about these things?
- Infrastructure. If privacy is an
- 5 important value for us as a society then we need
- to invest in infrastructure to support it.
- 7 Concrete guidance. We actually need to
- ⁸ go beyond just systems should have use
- 9 limitations. We need to actually tell people how
- are you're going to do it? And I can get into
- that more if people have questions.
- But we need to be writing really
- specific guidance, rather than just the, you need
- to have notice and consent, you should be
- thinking about use limitations, things like that.
- And last, everything in the world can be
- datafied these days, including how these systems
- 18 are working and how effective they are. And we
- can do the analysis, and we can get beyond
- anecdotes, and we can start analyzing data and
- figuring out is this effective, is this not
- effective, is this having negative effects, is

- this creating bias analysis, etcetera.
- Thanks very much.
- MS. BRAND: Thank you. And our last
- 4 panelist is Chris Inglis. He is currently a
- ⁵ venture partner at Paladin Capital Group and is
- 6 the former Deputy Director of the NSA. Thanks
- ⁷ for being here.
- MR. INGLIS: Great, thank you. And I'm
- 9 honor bound to say that I spend most of my time
- teaching at the Naval Academy in the Computer
- 11 Science and the Cyber Operations Department.
- First, I, like the other panelists, am
- grateful that you've established this venue for
- 14 what I think is a really important dialogue, and
- 15 I'd like to make four quick comments and then
- help us get to question and answers.
- First and foremost, I absolutely agree
- with the panel's premise, which I believe is, is
- that the framers of the Constitution did not
- intend for security and privacy to be in mortal
- 21 combat and we're therefore trying to figure out
- how do we achieve both.

- And it may well be that we cannot trade
- one for the other. I think that's right, we
- 3 cannot, but we have to work harder to achieve
- 4 both. And I think technology and practice from
- 5 the private sector can be helpful there.
- Two, I agree that government is
- different, not simply in the powers, the tools
- 8 that it might bring to bear on its citizenry or
- others, and therefore should be constrained, but
- the government alone has the requirement to
- essentially meet the standards of the First,
- 12 Fourth and Tenth Amendments within the
- 13 Constitution.
- I will tell you that from my NSA
- experience, the Tenth Amendment was the most
- significant of those, which essentially says
- unless you have the authority to do something,
- you do not, you know.
- And against what has been said, which is
- that backdoor searches or 215 was NSA's
- interpretation, both of those were specifically
- 22 permitted under court approved procedures and

- specifically were interpretations of the law that
- went through three branches of government.
- I think that's right and proper. That
- 4 doesn't necessarily justify them. It may be bad
- ⁵ policy at the end of the day, but the rule of law
- has to pertain, right, in terms of how the
- ⁷ government gets things done.
- Point three, I would say that largely I
- 9 agree with what John had to say. Matter of fact,
- 10 I wholly agree with what John had to say, that
- the aspects of law and technology are often at
- odds with one another, not because they cannot be
- 13 reconciled, but because they're perceived as
- independent biases on any particular solution.
- And I would add a third, which is that
- what typically plays out in any one of these
- 17 systems is that you're trying to effect a
- technology, law, and the operational practice of
- those who essentially make use of the technology.
- 20 And the unsurprising result is that
- 21 because they do not change at the same rate, they
- essentially change at very different rates,

- keeping them reconciled or synchronized from
- 2 moment to moment is really hard.
- Therefore, mechanisms, FIPPs-like
- 4 mechanisms or other things are not likely to
- ⁵ satisfy the need.
- What you need are threads or systemic
- ⁷ solutions that essentially you pull through and
- you take both art and science process to
- 9 essentially try to figure out how to make some
- 10 solution here.
- 11 I'll wholly agree with John that
- education's going to be absolutely essential. At
- NSA ultimately when we found ourselves in the
- midst of some compliance incidents for which no
- one had intentionally made a mistake, we actually
- had to sit down and figure out how do you achieve
- ¹⁷ a horizontal join between the technologists, the
- legal practitioners and the operators, all of
- whom were trying to achieve something that was
- slightly different, but ultimately invested in
- the same problem set.
- The last point I would make is that I do

- believe that there's a role for big data, what is
- ² sometimes called mass collection. There's a role
- for big data, but the principles that should
- 4 pertain to the government's collection of that
- 5 should be the same as surgical data, which is
- 6 necessity and proportionality.
- 7 The government should be able to justify
- on what basis this is necessary, such that it
- 9 could then argue, not for an encroachment upon
- civil liberties or privacy, but how do we then
- work harder to achieve the sustainment of privacy
- 12 and civil liberty. And it should only achieve
- that in proportion to that need.
- Therefore, I think that all those four
- 15 comments aside, I would say that the private
- 16 sector probably has a lot of experience in this
- 17 regard that the government can take advantage of.
- My own sense is that the government
- 19 collects far less information than is perceived
- by the public, and certainly far less information
- than the private sector does.
- Again, I don't excuse the government for

- $^{
 m 1}$ that. The government should be held to account,
- but the government can, in fact, bring
- 3 technologies in that might well scale quite well
- for the government's purposes because we'd have
- to scale them down, as opposed to scale them up.
- I'm open to any questions you may have.
- MS. BRAND: Thank you. Just a reminder
- 8 to the audience that there are PCLOB staffers in
- ⁹ the back with cards and if you'd like to direct a
- written question to the panelists, hold up your
- hand, find one of them and then write down your
- 12 question.
- And for the benefit of the audience and
- the cameras, for the panelists, when you're
- answering a question if you wouldn't mind moving
- the mic back and forth. I'm sorry, we don't have
- as many mics as we probably should.
- So I'd like to start by asking about
- oversight, and I'd like, Mr. Grant, to direct
- this question to you first.
- Both in your oral statement and in the
- written statement that you submitted to us, you

- talked about a wide range of mechanisms, paper
- trails, electronic work flows and things like
- that, and frankly, on the written statement it
- 4 seemed like an overwhelming array of different
- ways to engage in oversight.
- I think for a couple of reasons you need
- ⁷ to choose your oversight mechanisms. One is that
- 8 any agency is going to have limited resources to
- ⁹ dedicate to oversight.
- And secondly, as I mentioned in a
- previous panel, there may come a point where
- there are diminishing returns on oversight. You
- need to leave these people doing the work at the
- NSA or other agencies time to actually do their
- job, not just comply with oversight mechanisms
- 16 all day long. So you have to find some balance.
- So have you given some thought to what
- 18 constitutes an effective oversight mechanism?
- 19 How do you rank different mechanisms in terms of
- their effectiveness?
- MR. GRANT: Yes, so I think we should
- 22 actually think about oversight as a big data

- 1 problem and then apply the same thinking to it
- that we would apply to trying to analyze signals
- intelligence and trying to analyze huge amounts
- ⁴ of transactional data for marketing.
- It's a similar issue. You have a huge
- 6 amount of data, as you say. There are massive
- amounts of audit logs, for example, in an
- 8 organization like the NSA, and that's a lot of
- ⁹ information.
- But you can use technology and analytic
- tools to make sense of that information and
- derive the insights that you're looking for.
- So but part of the issue is, A, you need
- to do it, you need someone. So we see this all
- the time in Palantir and I know other
- organizations see this as well, which is,
- everybody checks the box on FISMA for audit logs.
- We've got the audit logs and we will go through
- an enormous number of hoops to make sure it's
- logging exactly the information that it's
- supposed to.
- We get fewer requests to actually look

- at the audit logs once the auditing mechanisms
- ² are turned on.
- And looking back to my congressional
- 4 experience, there aren't many laws that I can
- 5 recall that tell anyone they actually have to
- look at the audit logs, they just have to.
- 7 It's the Seinfeld joke about renting a
- 8 car. Everybody can take the reservation but you
- 9 have to hold the reservation. You have to use
- the information.
- So I think, I mean to me that's how you
- make oversight more effective, you use these
- 13 techniques.
- And that's another thing. The oversight
- people and the information security people and
- things like that, they should be as good as your
- analysts, and you need to have good people who
- 18 are also doing the analysis and conducting the
- oversight.
- So to get to your last question, which
- is the most effective? I think it's using that
- 22 auditing data. I think it's using that big data

- that you've got and having a team of people that
- ² can proactively comb through it.
- And not only are you going to look for
- 4 people doing something wrong, but you can also
- ask questions such as, you know, does our data
- ferention policy make sense?
- You could look at the data and say, you
- 8 know what, it turns out we keep data, this data
- 9 set for five years. Nobody ever uses the data
- older than three years in that data set, so let's
- 11 change the data retention policy to fit with the
- 12 actual usage of the data.
- MS. BRAND: Thank you. Mr. Inglis, I'd
- especially like your thoughts from your time in
- government, what did you view as an effective
- oversight mechanism?
- MR. INGLIS: So first and foremost, if
- there is an authority that is granted or a burden
- that's imposed, and they come hand in glove, you
- 20 know, that's not a one time thing. There cannot
- be a repurposing somewhere later simply having
- gotten past that threshold.

- 1 At NSA the typical events might be
- ² constituted as collection, processing of data,
- analysis of data, dissemination of that data, and
- the burden was imposed at every step according to
- ⁵ whatever the authorities were that were granted
- for the acquiring of that data, the acquisition
- of that data in the first place.
- And what we ultimately found is that in
- ⁹ order to achieve that, because data ultimately is
- aggregated, synthesized, a typical
- counterterrorism analyst, we take the iconic
- analytic effort, doesn't simply use data from one
- source, they use data from many sources.
- And at that point it is really hard, if
- there are different expectations of the different
- data sets to try and keep it straight in your
- head as to what you're going to do about that.
- So the focus has to be how do you bind
- 19 the attributes for a particular data element at
- the moment that it comes into being?
- MS. BRAND: Could you pull the mic a
- 22 little closer.

1 MR. INGLIS: At the moment you collect a 2 piece of data, how do you bind the attributes to 3 that data that essentially include, but perhaps some other things as well, what was the authority 5 under which this data was collected? What are 6 the burdens? What are the imposed constraints 7 that come along with that? What are the 8 proscriptions, if any, associated with that? 9 And that should be atomically bound to 10 that data element through its life, through its 11 life of collection, processing, analysis and dissemination. 12 13 Now at some point there's going to be a 14 second order use of that data where someone 15 essentially reads a broad swath of material, 16 synthesizes that in their head and then 17 constructs a document across an air gap. 18 That gets hard, but at least in that 19 primary use of that data if you had a systemic 20 view of it from start to finish, you make the 21 auditor's job or the compliance oversight much, 22 much easier.

- And you therefore then in your system,
- in your technology essentially impose a
- 3 constraint or a check every time something
- exercises privilege against that data, whether
- ⁵ it's a collection, or analysis, or processing, or
- in dissemination. That makes the auditor's job
- ⁷ much easier.
- And frankly, it has a nice deterrent
- 9 effect on those inside the system because they
- 10 know at every moment that they are held to
- 11 account.
- But in my experience in government it's
- 13 not so much the deterrent as it is the assist in
- an otherwise very, very rule-laden environment.
- A typical counterterrorism analyst at
- NSA would often deal with hundreds of constraints
- 17 on the data sets that are available to them
- 18 because various orders of the court,
- 19 interpretations of the court, kind of sharing
- arrangements with various other nations would all
- 21 come along with their independent assessments of
- how the data can or should be used.

- So the bottom line is that technology
- ² can help us by essentially doing an atomic bind,
- ³ right, meaning that it's organic to the data
- 4 itself of what's its provenance, and that
- 5 provenance should never be lost through the
- 6 history of that system.
- MS. BRAND: Thank you. I'd like to turn
- 8 to the FIPPs, and Mr. Geiger, I was happy that
- ⁹ you raised those, and Professor Cate as well. So
- 10 I'd like to direct this question first to the two
- of you.
- So, Mr. Geiger, I noticed that in the
- written statement that you sent to us, you talked
- 14 about the FIPPs but you didn't really talk about
- the individual participation FIPP.
- And I guess when I talk about the FIPPs
- 17 I'm referring primarily to the DHS version.
- You said in your oral statement just now
- that the FIPPs are not a smorgasbord, they're a
- framework, you can't just pick and choose among
- them. And if that's the case and if you have to
- employ the individual participation FIPP, how can

- that work in a surveillance context?
- MR. GEIGER: So that is the toughest
- FIPP to apply in this context, absolutely.
- One way that you could do it, which is
- ⁵ not politically viable and perhaps not even good
- 6 policy, would be to loosen standing requirements
- on individuals to bring suit for violations of
- 8 law.
- 9 But my, I think, more reasoned answer is
- that if the individual participation FIPP is
- lacking in the national security context, then
- the rest of the framework has to work overtime to
- compensate.
- And that includes data minimization,
- which is why I emphasize collection limitations
- and transparency, as well as the rest of the
- ¹⁷ framework.
- And I mean I absolutely recognize the
- 19 challenges in applying individual participation,
- but this is one area, again, where government is
- different than the private sector and I think
- that difference should express itself in

- particular in the data minimization principle.
- MS. BRAND: Professor Cate, do you have
- 3 thoughts on that?
- 4 And I would ask also, there's been a lot
- written and said in public more recently about
- 6 how perhaps the consent and individual notice
- FIPP really doesn't work very well in the private
- 8 sector, either because nobody really understands
- ⁹ what they're consenting to. Even if they
- understand it they don't have any other option,
- so they have to consent to get the service, and
- it's kind of a meaningless exercise.
- Do you have thoughts on that and whether
- the individual participation FIPP can work in
- 15 this context?
- MR. CATE: Thank you very much. I do
- have thoughts on that, especially being one of
- the people who's written some of that.
- 19 I think the challenge of the FIPPs is
- that the they often lead us in the wrong
- direction. And I think this is a real challenge.
- 22 I'm not in any way trying to make it sound easy

- or make it sound like there's a simple answer
- ² here.
- But, for example, if we think about the
- ⁴ FIPPs in sort of their classic 1980s OECD FIPPs,
- we're talking about notice and consent, we're
- talking about purpose specification, we're
- ⁷ talking about use limitation to the purpose
- 8 specified, and then we add things like data
- 9 minimization and individual participation.
- And frankly, almost all of these seem
- challenged in a modern data environment, private
- 12 sector or public sector.
- In other words, how does that really
- work? You know, there are 60 people in the room,
- they all have cellphones, they have recording
- devices, they have video, they have audio, I
- don't have a policy statement from any of them.
- 18 I don't know about my individual participation
- 19 rights. I suspect they would look down on my
- wanting to interview them each about it.
- It's not a meaningful way to approach
- the issue. The issue is an important one, which

- is how to protect privacy. But shifting the
- burden to the individual, which is what the FIPPs
- have the larger effect of doing, is a very
- 4 difficult way to approach that. And I think it's
- 5 an impossible way to approach it in the public
- 6 sector environment.
- But it also may lead to completely wrong
- 8 results. In other words, one of the surprising
- ⁹ things to me, and I can't believe I'm going to
- say this in a place that's being recorded, but
- about the Section 215 was that the NSA collected
- 12 all this data and did so little with it. It was
- 13 astonishing.
- And so you would like to say, you know,
- when people talk about atomically binding the
- limits on what you can do with the data with the
- data, I'd like to think if we thought of
- something new we might do with the data that
- might really have a major effect on national
- security, we'd have a process for some sort of
- risk analysis, what's the benefit, what's the
- risk, what are the processes in place to protect

- it, now let's do that thing.
- In other words, data has real value. It
- does in the national security environment, it
- 4 does in the private sector environment.
- 5 And I think we need to be thinking about
- 6 approaches here that aren't binding everything to
- ⁷ some mythical transaction that took place at
- 8 which in the FIPPs world we say the individual
- 9 agreed to this, even though I can't think of a
- case in which the individual actually agreed to
- it or it was meaningful consent.
- 12 And then in the national security world
- we just overlook that. We just say, well, we've
- 14 agreed for the individual because we think it was
- important, without again doing a clear and well-
- documented type of risk assessment using clearly
- 17 articulated values and harms, benefits and harms.
- MS. BRAND: Go ahead.
- MR. GEIGER: If I can just make three
- additional comments on the FIPPs. One, so it
- does sometimes lead programs in the wrong
- ²² direction.

- 1 It is a useful framework for evaluating
- ² privacy protection, but the application of the
- FIPPs, what you're actually doing with the
- 4 program, you may pass muster under your privacy
- impact assessment, but the actual the way the
- 6 program is conducted on the grounds may not in
- ⁷ fact be privacy protective.
- 8 So I don't think that the FIPPs are a
- ⁹ silver bullet, but the principles themselves I
- think are very useful for the evaluation of the
- 11 program.
- Second, there's been a long-standing
- 13 controversy about notice and consent being
- inadequate, but that is why I said at the outset
- that the FIPPs is a framework. I mean each
- principle is dependent on the other.
- This came up very clearly in the
- 18 healthcare context. People don't know what
- they're consenting to when they receive a notice
- from their doctor. They don't know what the
- 21 HIPAA privacy notice really says or means, or
- what HIPAA does, which is why there has to be a

- 1 lot of additional privacy protections in place to
- ² actually meaningfully protect that individual's
- ³ privacy.
- And then lastly, FIPPs are not the only
- ⁵ framework. I think that it is a very useful, I
- think it's an indispensable framework, but there
- ⁷ are certainly other frameworks that can be
- 8 applied and should be applied to the evaluation
- 9 of security or data collection programs writ
- ¹⁰ large.
- MS. BRAND: Although this was the
- 12 subject of the first panel today and not
- 13 necessarily this panel, I want to ask about it
- anyway. So apologies if I'm springing this on
- ¹⁵ you.
- But I'd like to give you all a chance to
- give any views you might have on privacy, what is
- 18 privacy, the sort of nature of the underlying
- ¹⁹ privacy right.
- 20 And Mr. Inglis in particular, when you
- were at the NSA, I assume you spent some of your
- time thinking about how to protect privacy and

- civil liberties and as you were doing that, what
- did you think that meant? What privacy interests
- were you trying to protect?
- 4 MR. INGLIS: I would say I don't think
- that has changed over time, though the technology
- 6 might hold that at risk in different ways and
- there might be some downstream consequences,
- given the scope and scale. But the fundamental
- 9 question always comes back to two things.
- One, with respect to the perspective of
- the individual, is there a reasonable expectation
- of privacy for, fill in the blank what that
- information might be.
- That's the stuff of great legal debate,
- but operators think about that as well,
- particularly the operators inside the government
- because they're constrained by the Tenth
- 18 Amendment to think about what is it they're
- 19 actually authorized to do, everything else then
- 20 being proscribed.
- But the second way to think about the
- issue of privacy is then what might you learn if

- 1 you take these discrete data sets and combine
- them in a way that might then give you some
- insight into things that were not self-evident
- 4 from any one of the discrete data sets.
- 5 You have to therefore think about the
- ⁶ problem in the aggregation, synthesis downstream.
- ⁷ Again, you might have some thresholds there that
- you have to think your way through that you don't
- ⁹ want to go beyond at that particular point in
- 10 time.
- I would tell you that at the National
- 12 Security Agency ethos is as important as the
- compliance rules, the FIPPs mechanisms and things
- 14 of that sort. Absent ethos, absent the art,
- right, the science will lead you astray and
- you'll essentially get into a place where science
- 17 alone cannot help you essentially navigate the
- challenge, the question of how do you achieve
- both security and privacy in a world where they
- are massively converged in a place called the
- ²¹ Internet.
- MS. BRAND: Professor Cate, do you have

- a thought on the nature of privacy?
- MR. CATE: I was afraid we might run out
- of time before you got to me on this.
- I would say two things. One, this is an
- 5 area where I think public sector versus private
- 6 sector is a really important distinction and I
- ⁷ think it has to be kept clearly in mind.
- In the private sector I think of privacy
- 9 mainly in terms of, if you will, harms or impacts
- on individuals or on groups of individuals.
- So whether that's the way we think about
- it in the Fair Credit Reporting Act, like a
- 13 higher price for credit or denying someone a
- benefit, or whether it's some other way in which
- we think about an individual being manipulated or
- being driven to pay a higher price or what have
- ¹⁷ you.
- In the public sector I think that is
- 19 also true. I think all those specific impacts,
- those harms, if you will, although I don't mean
- to limit them to physical or financial harms, are
- 22 present as well.

- But I think there's probably something
- more in the public sector, which is privacy, from
- 3 I think the very beginning of the constitutional
- debate, was seen as something about the balance
- of power between the individuals and their
- 6 government, between the citizenry and the
- ⁷ government.
- There is something quite striking, and
- ⁹ this I completely agree with Harley about, the
- more the government knows about individuals, the
- 11 greater the risk that that information will be
- used in a way that alters that balance of power,
- that makes the government more powerful and makes
- the individual less powerful.
- And it's, you know, a widely observed
- but an ironic twist as we've gone into the
- twenty-first century, we've in many ways gotten
- 18 less transparency to the citizen about the
- 19 government and more transparency about the
- 20 citizen to the government.
- That is a clear alteration in that
- relationship, that power relationship or that

- oversight relationship.
- So in that sense that's why, again,
- whether one focuses on collection or use it may
- be a not so significant matter, but I think at
- the end of the day it is use that matters. It's
- 6 knowing how can the government use this
- ⁷ information in a way that might affect me, as
- 8 opposed to is the information out there, which
- 9 seems to almost be the answer is yes now.
- MS. BRAND: Mr. Grant.
- MR. GRANT: I don't have necessarily the
- answer, but I think I have sort of a framework
- for thinking about it, which is to start to think
- about it from the perspective of social media
- 15 right now because I think in that space you're
- seeing how, especially younger people, are
- viewing privacy.
- 18 If you ask, so most of the engineers at
- 19 Palantir are, they appear to be about 14, and we
- 20 had some discussions internally about sort of our
- own information security policies and should the
- company be able to look at social media like

- 1 LinkedIn, Facebook, publicly available
- information, but look at it as part of our own
- inside policy. There are ways to detect phishing
- 4 and things like that using this kind of data.
- 5 And they vigorously objected to their
- own employer looking at that data, again, for a
- ⁷ reason of information security.
- 8 So it was interesting to explore with
- ⁹ them and to say, but you tweeted, you tweeted
- that, which means people are going to read that.
- 11 It is a tool for communication to the world.
- 12 And they still felt, yeah, it is
- 13 publicly available, anybody can Google it, but
- they still have an objection to government
- collecting it, or even government reading it, and
- then their employer reading it and things like
- 17 that.
- So I don't know exactly what that means
- in terms of coming up with a final definition of
- 20 privacy, but it suggests that people, there is a
- 21 different view of it. And that even public
- information, there's still privacy inherent in

- 1 public information somehow.
- And like I said, I think talking through
- 3 sort of attitudes towards social media and
- understanding that could help us figure out what
- is this, the newer conception of privacy in this
- 6 technological age.
- MS. BRAND: Did you have something to
- 8 say Mr. Geiger?
- 9 MR. GEIGER: Sure. I mean, I said most
- of it during my opening remark. I mean I do view
- privacy in the lens of control. I view it as an
- individual's ability to control information about
- herself, but then also the control that the
- entity holding information can exercise over
- ¹⁵ individuals.
- I think it is very important not to just
- 17 look at privacy harms, or privacy interests, or
- the extent that privacy can translate to control
- over a individual or their decisions in the
- context of today's technology.
- I think that it's very important to try
- to look out the next couple of decades and sort

- of see what is coming down the pike. And there
- ² are some very pervasive, very privacy intrusive
- ³ technologies that are, that I think we will see
- 4 in our homes and maybe even in ourselves in our
- ⁵ lifetimes and certainly in our children's
- 6 lifetimes.
- And the laws have absolutely not kept
- pace, and without a change in the law, again, I
- ⁹ reiterate that internal protections on use and
- 10 access, while important, are not sufficient
- because they can change. They have changed.
- When we talk about protecting privacy, I
- think that we should be looking, as I said, to
- what we are protecting several generations down
- 15 the line.
- MS. BRAND: Just to get back to the
- topic of this panel again, Professor Cate brought
- up use restrictions. We've been talking about
- 19 that throughout.
- We're focusing on how the private sector
- might have solutions that the government might
- learn from. Private companies are obviously

- doing something to control the use of information
- they collect. They have to. They have a privacy
- policy that says what they're going to do with
- 4 your information and they have to comply with it.
- 5 Are there mechanisms that the private
- 6 sectors has used for enforcing their use
- ⁷ limitations that are particularly effective that
- 8 the government might learn from?
- Mr. Grant, do you have a view on that?
- MR. GRANT: So we see this a lot
- obviously in terms of, so we, ourselves, don't
- 12 hold data but our customers hold data, and trying
- to help them implement compliance.
- Honestly, actually, they use the same
- basic mechanisms that I described in my testimony
- and often they have the same basic weaknesses.
- You know, do they have the
- infrastructure to manage access control to the
- 19 granular level? A lot of them do not because it
- 20 costs money and it takes time.
- 21 Are they conducting the oversight of the
- data? Probably more so than some people and

- possibly the government, again, because of
- limited resources. But they're still probably
- not doing it at the level that you would hope.
- One thing I notice is that a lot of
- them, there is, even in Europe where you have
- 6 more commercial privacy law and more commercial
- ⁷ privacy compliance requirements, a lot of times
- 8 it's best quess.
- So, for example, one that we've been
- 10 running into recently now is looking into
- 11 cybersecurity and information security, data
- exfiltration risks in the private sector.
- And in these giant, multinational
- companies they're trying to deal with employee
- privacy laws that are all over the map.
- And they're asking questions like, if a
- German employee sends an email to a U.S.
- employee, what privacy rules apply to the content
- of that email?
- In Germany you have to actually tell the
- employee, I'm about to start monitoring your
- email. In the United States you can pretty much

- 1 do whatever you want with a few exceptions. They
- don't know what the answer is so they make their
- 3 best quess.
- So I think there are interesting lessons
- in terms of what the privacy is trying to do, but
- 6 I actually think they're facing a lot of similar
- ⁷ problems that are related to scale, that are
- 8 related to lack of understanding of what the
- ⁹ rules should be, as the government.
- MS. BRAND: Anyone else have a thought
- on that question?
- MR. INGLIS: If I can add to it. So my
- own sense is that there's probably a lot of great
- technology out there that can be used, but any
- technology can fall short of your expectations if
- you don't use it in the right process, and
- therefore, we ought to give as much time and
- 18 attention to process within which that
- 19 technology might be used as the technology
- ²⁰ itself.
- In the following process it might be
- useful to consider that first and foremost before

- you acquire any capability, whether it's in the
- government or within the private sector, you
- 3 think your way through the necessity
- 4 proportionality considerations, you know, is this
- ⁵ necessary and have I done this only to the degree
- 6 that it is necessary.
- And then what we're trying to achieve is
- 8 not simply the balance between security and
- ⁹ privacy, but transparency is the third leg of
- that stool. And absent transparency, you often
- find yourself in a place where people don't
- believe that you achieved the right balance of
- the first two.
- 14 That then derives, you know, the
- possibility in the government the need to
- 16 essentially acquire explicit authority, which
- 17 always comes with constraints, constraints are
- bound to that, and some measure of accountability
- 19 for those constraints.
- The process elements that then are
- essentially implemented to pull that off, I think
- should have the aspect of continuous compliance,

- not discrete compliance at various phase points,
- but continuous compliance. You think about it
- 3 all the time, first, middle and last.
- 4 Kind of a stretched analogy is part of
- the problem with the absence of cybersecurity in
- 6 so many environments is you think about that as a
- ⁷ bolt-on. Until such time as we build our
- 8 systems, operate our systems continuously with
- ⁹ that foremost in mind as the primary attribute,
- it'll break our hearts.
- 11 The second process element of
- implementation is an external component.
- 13 Internal components are really essential. You
- 14 have to hold the people accountable internal with
- 15 the system.
- But unless there's an externally imposed
- accountability mechanism, you can wind up with
- mismatched expectations or the system might, in
- 19 fact, go roque.
- And then three, there has to be at
- various phase points required reporting, which is
- 22 important because that then forces some

- 1 synthesis, some kind of retrospective that says,
- how do we actually aggregate our experience in
- 3 this to come to some conclusions.
- So is it meeting our expectations? Is
- it working as it should? Are we a little bit
- for right of the course, left of the course such that
- we actually need to invest some time and energy
- in the process itself?
- Absent that, you find that you're the
- frog in the beaker and it's just getting a degree
- 11 hotter moment by moment, all of a sudden you're
- the boiled frog. And you hadn't realized because
- you didn't step back and take hard look at it
- that you actually got off course a little bit
- some time ago.
- MS. BRAND: Thank you, and I think my
- time is up, so we'll start with Mr. Dempsey and
- 18 go down the line.
- MR. DEMPSEY: Thank you, and thank you
- to the members of the panel for giving us your
- 21 time today.
- In a way building off of something that

- 1 Chris Inglis said, or at least that I heard you
- saying that we need the technology controls, we
- need to build the technology in a way that
- 4 implements these controls, but at the same time
- ⁵ we need policies that surround them. You need
- 6 the legal rules, etcetera.
- I think, John Grant, my first question
- 8 to you, you talked a lot about the potential in
- ⁹ terms of tagging information, and audit controls,
- and permission controls are very granular, but
- just to state the obvious, that's not a
- 12 substitute for legal rules and policies.
- MR. GRANT: Absolutely not. We try to
- say, you know, even when we talk about our
- privacy enhancing capabilities and stuff, if you
- think you're buying a switch that you can flick
- that protects privacy, it's not going to happen.
- 18 It's not possible.
- You have to be able to respond
- 20 dynamically to changing situations. You have to
- be able to make human-driven nuance decisions
- about data and about how it's used and is it

- being used appropriately. And that's just
- 2 something machines can't do.
- And it's the same reason we argue at
- 4 Palantir that you can't build a find terrorist
- button, that you need a human at the top of the
- 6 decision-making chain and at the top of the
- ⁷ analysis chain to do it.
- 8 And so I distrust any technology that
- 9 says don't worry about it, we've got privacy
- covered. And so what the goal should be for
- technologists is, what kinds of tools do policy
- makers need and then the oversight officers, the
- oversight boards, and the civil liberties
- 14 protection officers and things like that, what do
- they need and what makes their job easier or
- possible, especially when you're dealing with
- ¹⁷ data at scale.
- So an easy example is there's a lot
- work, a lot of research going into improving
- access control interface. When you're dealing,
- with terabytes of information in the
- 22 cybersecurities space, how can you create

- technological shortcuts to allow a human to make
- the decisions about how to manage that data?
- And that's how you do it. You think
- ⁴ about how do you support the policy, not how do
- ⁵ you replace the policy.
- 6 MR. DEMPSEY: Let me go to Fred. Fred,
- ⁷ totally accepting your point about the
- 8 limitations of the FIPPs and totally accepting
- your point about the importance of focusing on
- risk and focusing on use, you're not saying that
- 11 collection is irrelevant, that obviously the
- 12 Fourth Amendment is in some ways a collection
- 13 limitation.
- And that, you know, in a commercial
- context that company that had the flashlight app
- that was out collecting data, nobody even got to
- the harms analysis, that collection was
- inappropriate in and of itself.
- MR. CATE: Right. You are absolutely
- right and I agree completely. In other words,
- ²¹ I'm not suggesting collection is irrelevant, I'm
- suggesting we've made collection too much of the

- end of the story, so that once you cross, you
- know, it's like a spillway in a dam, once you're
- over the collection limit, then anything else
- 4 goes.
- MR. DEMPSEY: Well, the ironic thing is
- that at NSA, as Chris Inglis said, their view is
- ⁷ they never thought of it that way, that they
- 8 thought that you have your collection
- ⁹ authorization which is critical, your retention,
- your use, your dissemination, your retention
- 11 limit, that each one of those --
- MR. CATE: But if I can just respond to
- 13 that. I think there's something of a mismatch
- 14 here. And I'm not in any way doubting either
- what NSA was doing or what Chris is saying.
- But one of the astonishing things, for
- example, when I read the Section 215 report that
- came out from the NSA's civil liberties office, a
- well-written report, it was full of all of the
- limits on what they were doing and the incredible
- what can only be described as bureaucracy around
- that, both technical bureaucracy and human

- ¹ bureaucracy.
- But it sort of ignored the fact, which
- is what I think has struck most of the American
- 4 people, is how was the authorization obtained in
- 5 the first place?
- You know, we had a law that said
- ⁷ relevant to a specific investigation, you know,
- ⁸ 99 out of 100 people through relevant to a
- 9 specific investigation meant, might be focused on
- 10 specific individuals.
- 11 Apparently the 1 out of 100 who didn't
- was a FISA judge, and then had other judges there
- 13 along with him, and apparently some members of
- 14 Congress.
- So I think one of the critical issues
- when thinking about going forward is if this were
- the private sector there would have been
- 18 immediate customer feedback.
- You know, if that were Facebook
- interpreting that to say, by the way, you know
- under that privacy policy that says we'll only
- collect data for limited purposes, it means that

- we're going to collect absolutely everything, and
- then there would be customer reaction.
- What do we create that will mimic that
- in the classified environment, in the
- intelligence environment? Maybe that's the
- 6 PCLOB. I mean maybe that's literally having the
- outside of the agency but focused on privacy and
- 8 civil liberties that says we understand the
- 9 challenge but we think you've got the wrong end
- of the stick.
- But I think it is being overly focused,
- 12 for example, on the Fourth Amendment that creates
- this problem. As you well know, the FISC just
- dismissed the Fourth Amendment issues by saying,
- well, third-party doctrine, there's no problem at
- 16 all. Let's go ahead.
- And somebody should have been saying,
- wait a minute, you're talking about collecting
- data on everybody. And then that would have
- focused the discussion in a way that all of the
- technological controls and all of the
- 22 bureaucratic controls that have been now well-

- documented in the agency, somehow never did.
- MR. DEMPSEY: That's very helpful. I
- don't want to further rehash 215, the history of
- 4 215, and anyhow I have a red card so I guess
- 5 that's the end.
- 6 MR. MEDINE: So let me just follow-up
- ⁷ quickly on that point. Maybe what we need to do
- is supplement the FIPPs with the OMG standard,
- ⁹ which is, you know, in private practice I would
- have a client and I'd say, everything you've
- proposed to do is perfectly legal, but are you
- ¹² nuts?
- I mean how do we embed that stepping
- back and saying, okay, the lawyers have
- technically signed off, everyone has technically
- signed off, but this is a crazy thing to be
- doing?
- MR. CATE: Well, I mean I think one
- 19 positive step is adding someone like Becky
- Richards and an office to support her within the
- 21 agency. I think that's one way.
- So you get people who aren't just

- thinking about the law, but rather people who
- will say, I understand legal clearance is taken
- 3 care of, but I still have the oh, my God
- 4 response.
- 5 Are you allowed to refer to God at a
- 6 PCLOB hearing?
- 7 MR. MEDINE: It's free speech.
- 8 MR. CATE: I feel very nervous about
- ⁹ that.
- I think the PCLOB is another way. In
- other words, you say we're going to have some of
- those similar roles, not by any means identical,
- but outside of the agency.
- I think this is where I would say,
- although this just may reflect my naivete, you
- know, I would like to think that although we
- certainly need to have secret operations, we
- wouldn't have secret law.
- And so if a law that said one thing was
- being interpreted to mean the opposite, that
- someone would feel the need to signal that, as
- opposed to going out of their way to continue to

- say, no, it doesn't mean what we actually think
- it means, and it means what only you think it
- 3 means.
- 4 And so that we would build in avenues
- ⁵ for transparency about the law, so that at least
- 6 we all knew what the rules were going into it.
- And I think that's a huge problem when
- 8 the law itself is effectively classified because
- ⁹ of the way in which the interpretative process
- works.
- MR. MEDINE: Sure, John.
- MR. GRANT: Can I just jump in on that?
- 13 How we embed that in the private sector, or
- 14 certainly in our company, and it goes back to my
- point about education.
- Engineers and technologists think of
- things in terms of does it work or does it not
- work, and they just want to make things more
- 19 efficient.
- But it's not because they don't care
- 21 about privacy and civil liberties. They end up
- living in the world they create. It's just they

- don't realize that this raises an issue.
- So if you improve education across the
- board so that the technologists throughout the
- 4 NSA and throughout the private sector that are
- ⁵ building the capabilities and things like that,
- if they're all conscious of privacy and civil
- liberties, they're going to raise these questions
- 8 too. They're going to say, what are we building?
- And especially technology is an
- interesting place because it's the place where
- the engineer, the lowly engineer is more powerful
- than the CEO, because if the engineer says, I'm
- 13 not going to build this, then that's it. And if
- the CEO says, I'm going to fire you, they say,
- okay, I've got four more job offers to go
- somewhere else. So there's a really interesting
- power imbalance there within the organization.
- So if you instill the values that you're
- 19 looking for throughout the organization in the
- people, that's where you're going to get the OMG
- response.
- MR. MEDINE: I have a question for

- 1 Harley and Chris. In our 702 report we noted
- that most of the information that was collected
- wasn't reviewed and therefore wasn't minimized,
- 4 and that even of the information that was
- 5 collected oftentimes it wasn't minimized in terms
- of being deleted because there wasn't a
- determination about whether it had foreign
- 8 intelligence value. Harley proposes doing the
- 9 minimization up-front when it comes in.
- So I have a question for each of you.
- One is, Harley, is that a practical matter given
- how much information is coming in?
- And I guess to Chris, if that's not a
- 14 practical way, how do we do minimization better?
- MR. GEIGER: First, an unsolicited
- answer to your first question, which is in
- 17 addition to the proposals that have just been
- discussed I think a FISA court special advocate
- would also help with the OMG standard.
- I think that it's a multi-layered
- solution having privacy and civil liberties
- offices in agencies, a PCLOB and a FISA court

- special advocate hopefully gets us there.
- In terms of whether front-end, so what I
- 3 had said was that front-end minimization and
- back-end minimization are important. And so I
- 5 actually, one of the things that I had said was
- that the determination ought to be made whether
- ⁷ the information was needed and then flush it as a
- 8 default unless that determination is made.
- This is different than the way that I
- think it's done, at least in some agencies where
- they keep the information unless they make a
- determination that they don't need it, which is
- very different. And that sometimes causes
- information to languish. I think that that
- should be flipped.
- In terms of front-end information data
- collection, I do think that it can be feasible,
- but it also depends on the specific program, it
- depends on the purpose.
- And if the purpose is we're going to
- 21 collect everything, that sets off the OMG
- 22 standard for me.

- But if the purpose is narrower, and I
- think generally speaking it should be, then yes,
- there should be data collection limitations.
- I understand that there are technical
- ⁵ limitations there and that depending on the
- 6 actual means of data collection, sometimes it may
- be unavoidable that you collect more than you
- need, but then you should be flushing the
- 9 information that you don't need.
- MR. MEDINE: I'm probably going to run
- out of time, so Chris, if you have any reactions
- 12 to that?
- MR. INGLIS: Yes, so on both parts, so
- the question of 215, I know we don't really want
- to rehash whether that's good or bad policy, but
- from an NSA perspective three branches of
- government participated in the creation of that
- program, sustainment of that over years time,
- multiple administrations, more than three dozen
- ²⁰ judges.
- And so from an NSA perspective, charged
- to essentially effect the will of government,

- short of a referendum amongst 315 million people,
- which we do every two years, I don't know how you
- 3 actually kind of make a significant change in
- 4 terms of how the government comes to some of
- ⁵ those conclusions.
- The PCLOB is an extremely valuable
- ⁷ addition, but you know, I think that we're always
- ⁸ going to find ourselves in a place where
- ⁹ stakeholders stand in the shoes of those they
- 10 serve.
- With respect to your specific question,
- it's problematic on a couple of counts. You
- 13 know, first and foremost, if you try to minimize
- 14 at the point of collection you then ironically,
- paradoxically begin to focus on things that you
- shouldn't.
- The strange truth in the world is that
- there are two ends of every communication in the
- world, sometimes more, right, if you add in the
- courtesy copies and the blind courtesy copies.
- 21 And if your interest is legitimately in
- party one and you begin to then focus on party

- two, right, who is involved in that conversation,
- without merit, without some reasonable or
- probable cause, you then begin to encroach upon
- 4 their expectation of privacy, absent some kind of
- ⁵ reason to do so.
- So the policy at this moment essentially
- ⁷ uses this, upon recognition, which isn't a sloppy
- 8 policy. It just says do not focus undue
- ⁹ attention on that, and when you do encounter
- someone who deserves further protection, take it.
- 11 You must take it.
- Built into that then are some time
- 13 limitations for how long you can hold that data,
- and some necessity and proportionality conditions
- that say how much data is enough, for what
- purpose, and how long are you going to keep that
- without some meritorious reason.
- So if it participates or contributes to
- a report you keep that for longer. If it
- doesn't, then there are time limitations, you age
- it off. And those are always prescribed by those
- who essentially grant us our authority.

- MR. MEDINE: Thank you.
- MS. BRAND: Ms. Cook.
- MS. COLLINS COOK: So following up
- actually on a phrase that's been used a number of
- times today and asking the same question I asked
- a previous panel, there's this notion of
- ⁷ reasonable expectation of privacy.
- 8 To the extent that that evolves over
- ⁹ time, which I think that it does, how does one
- ascertain what is a reasonable expectation of
- 11 privacy?
- 12 Is it based on a Washington Post poll
- that 50 percent of Americans are uncomfortable
- with X, Y or Z? Is it the conduct that
- individuals nonetheless engage in, that they're
- uncomfortable about communications surveillance
- but people still use their phones, they still
- 18 engage in the world?
- 19 If we were going to look to reasonable
- expectation of privacy as a touchstone, how
- should we ascertain what it is?
- This is a question for the panel.

- 1 Chris, I'll start with you because you had
- indicated that the NSA did look to reasonable
- expectation of privacy as one of their
- ⁴ quidelines.
- MR. INGLIS: First and foremost, there's
- a basis of law which doesn't, if the technology
- 7 changes over time give us, say, a free pass to
- 8 say because the law allowed us to use the old
- ⁹ technology in this way, the new technology, which
- is more intrusive, can simply just continue
- unabated.
- But there is a wide practice of law and
- the NSA considers that, you know, as it makes its
- 14 appeals for authorities, which are always
- conditioned upon a Department of Justice
- 16 representation and the right authority, either
- under 12333 or the courts.
- Second, there is an expectation at a
- 19 place like NSA that you think through the eyes of
- those whose privacy would be encroached upon,
- right. So you think about what's the expectation
- of the individual and is their expectation such,

- 1 regardless of what the law might say, that this
- is something that deserves some aspect of
- ³ privacy.
- 4 And that necessarily then has to inform
- the conversation about what authorities you seek
- and what then provisions you seek those
- ⁷ authorities for.
- 8 Interesting dialogue earlier about the
- ⁹ 215 program and the internal bureaucracy. At NSA
- we thought that was a feature, right, that the
- 11 court essentially proscribed use of that database
- 12 for anything but the very surgical and narrow
- ¹³ application of it.
- The sense at NSA was, is that if we had
- even requested to use that for other purposes,
- say, domestic terrorism, which is not our
- provenance, or say, weapons of mass destruction,
- 18 roque nations, that that would have been an
- encroachment into privacy that was excessive and
- therefore not meritorious right up front with
- respect to the possibility we might ask for that.
- The program as designed was very

- surgically, narrowly framed on something alone,
- which was warranted and justified under the
- 3 concept of necessity and proportionality.
- And we had to avoid the creep beyond
- 5 that because of an expectation based upon the
- 6 consumer looking back at us, as to what they
- ⁷ might think.
- MS. COLLINS COOK: John, you also used
- ⁹ the phrase as well I think here, and so if you
- have some thoughts on this.
- MR. GRANT: The thing that jumps to my
- mind and it gets back into, again, when I was
- 13 talking about analyzing data to sort of support
- the effects of this program, I think it's
- 15 reasonable to expect that the government won't
- look at data that's not useful.
- 17 That is a reasonable expectation of
- privacy, that the information that has not proven
- 19 effective for some purpose, that that won't be
- 20 collected and analyzed.
- And that's what we've been doing, as I
- said, rewriting our internal information and

- security policies, and as we've surveyed
- everybody at the company they've said, I'm fine
- with you looking at some of this data, just tell
- 4 me that it's useful, tell me why you're looking
- 5 at it.
- Because of course they're interested in
- ⁷ protecting our own internal information security
- 8 at Palantir, and of course we're interested in
- 9 protecting our own national security.
- So I mean this isn't the only standard.
- 11 Obviously utility can't be the only analysis
- point because there obviously are interests
- beyond that, but I think it's a significant
- question that we don't answer very well right
- 15 now.
- And this is, you know again, across the
- board from that sector to the private sector,
- everybody wants data and they think they can do
- all of this stuff with data.
- And we get customers all the time who'll
- come in and say, I've got to understand the
- Twitter. And we'll say, well, what do you want

- to know? And half the time, we'll say that
- information, if you want to understand do a lot
- of people like Justin Bieber or cats, then
- ⁴ Twitter's great. If you want to understand more
- 5 complex, nuanced theory, then maybe we should
- 6 think about something else.
- And I think that government should do
- 8 the same. And I think the government can answer
- ⁹ those questions, again, looking at, analyzing how
- data is used and using that data about data.
- So to me that's one area where you would
- sort of expand that definition of reasonable
- expectation of privacy, which is it's reasonable
- to expect no one will look at data that isn't
- 15 useful.
- MR. GEIGER: The question you pose is a
- very difficult one. I mean courts are wrestling
- with it all the time. And everyone has a
- personal opinion about it, and so do I. I
- believe that reasonable expectation of privacy is
- 21 a terrible framework actually.
- The Fourth Amendment is supposed to

- 1 protect against unreasonable searches and
- ² seizures. The reasonable expectation of privacy
- is a judicial-made creation that has now allowed
- 4 for some very unreasonable searches and seizures.
- ⁵ Section 215 is a great example of that.
- Under the reasonable expectation of
- 7 privacy framework, U.S. versus Jones
- 8 notwithstanding, because I know that's kind of a
- 9 mysterious opinion, but the Supreme Court seems
- to be sort of moving, inching along perhaps in a
- direction where they are doubting the reasonable
- expectation of privacy framework as it's been
- ¹³ applied in the past several decades.
- But under current law would it be okay
- under the reasonable expectation of privacy test
- to have a network of drones or a network of
- 17 ground-based cameras that watch everything that
- you do the moment you step outside of your house?
- I mean there is a very strong argument
- that, yes, that is okay under the reasonable
- 21 expectation of privacy framework.
- So I think that it's the wrong framework

- to be viewing a lot of this stuff. I think that
- it does not have to be left out of the
- 3 conversation, just like the FIPPs, it is one
- 4 framework.
- 5 There should be multiple lenses, but
- 6 none of them, including reasonable expectation of
- ⁷ privacy, like the FIPPs, are going to be a silver
- bullet. And they're not going to provide you
- ⁹ with a clear answer.
- MS. COLLINS COOK: I think if I have
- time for one additional question, I'm still
- 12 seeing yellow.
- So moving up the analysis of data and
- 14 requiring agents or analysts to make an
- assessment of whether or not information is
- relevant or is necessary to maintain, rather than
- potentially letting that information simply age
- off of your system, what about the privacy
- implications of that type of approach, which to
- me, I have been unable to get past this notion
- that that would require agents or analysts to put
- eyes on more communications than they would

- otherwise review. And so what is your answer to
- the privacy implications of that shift?
- MR. GEIGER: I mean I suppose that there
- 4 are two ways to do it. You could require the
- 5 agent to look over every piece of data that
- 6 they've collected.
- If the amount of data is small, which is
- 8 my main point, I mean having, not data retention
- ⁹ but collection limitation at the front-end. If
- the data population is small that is less of a
- 11 problem.
- 12 If you're requiring the agents to look
- 13 through a large amount of data that you know
- 14 contains information about individuals who are
- not connected to a crime or terrorism, that
- becomes more of a problem.
- Then on the flip side I suppose you
- 18 could have the agent merely looking at data that
- 19 they know is connected to other parts of their
- work.
- I mean I don't think there's a hard and
- fast rule. It's going to be depend on the

- 1 program, it's going to depend on what the agent
- ² is looking for.
- For that reason I think that data
- minimization, again, on the back-end is not the
- 5 answer. It has to be part of the framework. And
- 6 collection limitation at the front-end is a
- ⁷ crucial part of that framework.
- MS. BRAND: Judge Wald.
- MS. WALD: Whether or not you think that
- it's important to limit collection or you think
- 11 perhaps you can wait a while or see and go after
- it more forcefully at the use end, I'm interested
- in what you think the role of the courts are.
- In our other systems like criminal
- justice, ultimately, and even under the Fourth
- 16 Amendment, the courts are kind of the final
- 17 analysis. And even in many of our civil
- 18 regulatory systems ultimately they come up.
- So the question is two parts. At what
- stage, whether you believe in collection
- limitations or you believe more in use, do you
- think the internal, all of the internal audits

- and various other techniques that we've talked
- about are not enough, that you need some kind of
- 3 an outside look at it?
- But secondly, I think as a former judge
- I ask this question, if you were scared to come
- before -- and that is, do you really think that
- ⁷ the limited role that the FISA court has been
- 8 allowed to play in terms of the secrecy of its
- 9 operations, and even with our recommendation and
- other people's suggestion about adding an
- adversary, and even some of the judges on that
- court, not only did they come out in different
- ways, all judges do, but they were frustrated
- themselves in terms of the technology sometimes.
- Judge Bates remarked that it was
- practically impossible, given all of the
- 17 complexity of the technology we've talked about
- and the fact that these judges would come in from
- their regular work for a week at a time and then
- go back again, is that the best kind of outside,
- 21 not outside surveillance, outside look, an
- independent look, or is there some better way to

- get the notion of an independent, the Supreme
- ² Court always talks about independent and neutral?
- It's a big question. Go at it, starting
- with Professor Cate, any way you want.
- MR. CATE: Thank you very much, Judge
- 6 Wald. I would say I think the role of the courts
- is absolutely essential. I think the important
- 8 feature of that role is it needs to be an
- ⁹ independent role, and I think one of the concerns
- with the FISC is that as this set of opinions
- went back and forth and, you know, small
- modifications, and updates, and briefings and
- corrections, it involved the court in the more
- daily operation of the agency than I think we
- would typically think appropriate or desirable,
- that we really want an independent, neutral and
- ¹⁷ detached court.
- The challenge of technology is huge for
- all of us. Even engineers have difficulty
- 20 keeping up with the technology. I think there
- 21 are, and we have seen some ways of dealing with
- it. One is court-appointed experts.

- 1 Another, as we saw in the Supreme
- ² Court's most recent privacy opinion this summer
- 3 it cited heavily to amicus briefs from CDT, and
- 4 from EPIC and others where they explained the
- technology and the impact of the technology, and
- the court clearly relied on them. And I think we
- 5 shouldn't overlook that.
- And then of course courts also have
- 9 remarkable powers to compel the parties to
- explain the technologies in clear and
- understandable language and to not accept their
- 12 filings or to not rule on their filings until
- 13 they do.
- So I could say more but let me share the
- microphone.
- MR. GEIGER: I absolutely agree with
- everything Professor Cate just said. The courts
- 18 play a very crucial role in the oversight of
- ¹⁹ national security surveillance programs. I think
- that the court is constrained by a lot of
- 21 statutory limitations.
- I think we would welcome, at least the

- privacy advocacy community would welcome court
- oversight of minimization procedures and on the
- ³ ground controls on privacy.
- I know the court does some of that, but
- ⁵ I know that it is also limited to sometimes just
- 6 a certification.
- We have talked about having a special
- 8 advocate. I don't necessarily view that person
- 9 as an adversary because I think that in many
- cases the court and the government are also
- trying to protect privacy, they just maybe differ
- on the strength of that privacy protection. So I
- think that the special advocate could, in fact,
- 14 be an ally.
- But then also technical experts and
- amicus. One of the problems that we're seeing in
- the debate over bringing in amici or bringing in
- a special advocate is that there are some forces
- in the court, perhaps formerly of the court, who
- would like to see greater restriction placed on
- those parties, so that it is the FISA court that
- instead gets to decide what role and what access

- 1 to information these amici will play, which will
- ² severely undercut their effectiveness and their
- ability to help the court. So I would urge
- 4 resisting those calls.
- MS. WALD: Don't you think, this is a
- follow-up just on the point you made, don't you
- ⁷ think that in some cases, even the legal or even
- 8 possibly constitutional reasonableness of
- 9 something is dependent on understanding the
- technology of it?
- I mean I think Judge Bates felt that
- ¹² way --
- MR. GEIGER: Certainly.
- MS. WALD: In one of the cases that was
- declassified and put out that way.
- So you think that they are equipped to
- do that now, or do you think the advocate will
- 18 fill that role, or do you need more?
- MR. GEIGER: So I don't know enough
- about the judges to make a determination about
- their level of familiarity with technology.
- But I mean this technology that is

- being exploited in some of these instances can be
- extremely complicated, and so, no, I would not
- 3 imagine that most lawyers have that sort of
- 4 training and so I think that there -- I know that
- ⁵ the court already has powers to some extent. I
- think those should be loosened to bring in
- ⁷ technical experts as amici to explain this in as
- 8 clear a manner as possible, because I think
- you're absolutely right, technology does have a
- direct bearing on the rights that are being
- 11 manipulated.
- MR. GRANT: And so I'll just jump off
- 13 that one. I think that it's critical to have a
- 14 translator role for the court, someone to help in
- an unbiased way try to explain the technology.
- And you know, this isn't just an issue
- for the court, it's an issue for Congress. You
- 18 know, I was trying to write cybersecurity
- legislation before I left and one of the
- 20 challenges was you have to have a really complex
- technical debate and members are naturally going
- to be uncomfortable taking a strong stand when

- they're not a hundred percent sure what the
- technological considerations are. And the end
- ³ result is you sort of paralyze things.
- I think the critical question, so the
- 5 court role is vital and it's important that it
- takes time because, you know, by nature that
- ⁷ briefs out the issues and it helps you understand
- 8 things.
- The challenge is what are you doing in
- between. Because technology becomes ubiquitous
- even in a matter of months sometimes, and it
- starts to have a real effect on people's lives
- right away, and it's going to take 10, 15 years
- sometimes for the court to eventually settle on
- what they want to do.
- So what do you do in the meantime and
- how should people be guided? Should there be
- ethical limitations on what the private sector
- wants to do? Should the government figure out
- ways to sort of slow walk in technology? And
- what's the framework for making that decision and
- implementing that? I think that's the real

- ¹ challenge.
- MR. INGLIS: I largely agree with what's
- been said. I think that with respect to the role
- of the court neutral and detached is, I think,
- the right way with respect to their opinion on
- the efficacy of the policy or the government's
- ⁷ representation. But they have to have a solid,
- if not exquisite understanding of the technology,
- 9 and I would distinguish between the two.
- I think the role of an adversary and a
- technology expert at the court, you know, has
- great merit and would, I think, add to their
- ability to at least understand the technology.
- And we have to hedge our expectations,
- not because the government wouldn't want to
- 16 reform, but at NSA could be perhaps exhaustive
- about technology at some moment in time in its
- 18 presentation to the court, but at best it can
- only be illustrative as to where that technology
- is going to go. Nobody knows where the
- technology's going to go.
- 22 And the use of a certain technology,

- even if the technology doesn't change, change is
- in and of itself. People make different use of
- 3 technologies.
- 4 And so forecasting that is, I wouldn't
- say a fool's errand, but it's really hard.
- MS. BRAND: Thank you. We have a couple
- of public questions. We may only have time to
- get to one of them.
- 9 But Professor Cate, I think this is
- directed at you. It says, if you don't like the
- 11 FIPPs, what alternative do you suggest?
- MR. CATE: So first of all, to be clear,
- 13 I'm not saying I don't like the FIPPs, I just
- don't think the FIPPs are the be all and end all.
- And second of all, I suggested risk
- management as a pretty useful tool as a way of
- 17 identifying both potential negative impacts and
- 18 also beneficial impacts.
- And you know, one of the things we
- 20 haven't talked about is the value of the use of
- 21 data for national security or foreign
- intelligence gathering or whatever.

- 1 And one advantage of a risk management
- ² approach is it helps focus on both sides of that
- ³ equation. It helps drive towards specificity.
- 4 So if you ever want a documented decision that
- ⁵ reflects that analysis, it's one way to help
- 6 focus attention on it.
- And then as we identify those potential
- 8 harmful impacts, negative impacts, whatever we
- 9 want to call them, we can then look for tools
- that help minimize those impacts.
- So if the harmful impact is if you
- 12 collect all this data it might be stolen, we can
- 13 talk about security.
- 14 If we collect all this data and the fear
- is that the government might repurpose it for
- some other use, then we can talk about use
- 17 limitations that would help address that.
- But I think a great advantage of doing
- this is it makes clear in a way that the FIPPs do
- not, where should we be focusing our attention,
- whether we are academics, or the PCLOB or, you
- 22 know, with the process within an agency.

- MS. BRAND: Okay, thank you.
- ² Mr. Chairman.
- MR. MEDINE: Thanks again to the
- 4 speakers on this panel and all the panels
- throughout the day, as well as the audience
- 6 members who submitted questions.
- I think we've had a remarkably
- informative and thoughtful discussion. We've
- 9 heard from academics, government officials,
- advocates, technologists in industry, which is a
- lot, and we've covered a broad range of topics,
- 12 FIPPs, Fourth Amendment, collection and use,
- encryption, de-identification, oversight,
- accountability, technology, mosaic theory and
- bulk data all in one day.
- So you've given us a lot to chew on. I
- think this is very helpful for us as we consider
- how to move forward carrying out our mission to
- balance national security with privacy and civil
- ²⁰ liberties.
- So unless any other Board members have
- 22 any comments, today's Board activities are

384 complete. 2 We encourage anyone who has comments, 3 whether panelists, or members of the audience, or others to submit written comments. 5 accepting comments on regulations.gov through the 6 end of the year. 7 A transcript, again, of this day's 8 activities will be posted on our website, 9 pclob.gov. 10 And with that, I move to adjourn the hearing. All in favor of adjourning say aye. 11 12 (Vote taken.) 13 MR. MEDINE: We are adjourned. It is 14 now 4:15. Thank you very much. (Whereupon, the hearing was adjourned.) 15 16 17 18 19 20 21 22

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                     CERTIFICATION
3
             I, LYNNE LIVINGSTON, A Notary Public of
5
    the State of Maryland, Baltimore County, do
6
    hereby certify that the proceedings contained
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14
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16
              Lynne Livingston
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              Notary Public
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              My commission expires: December 10,
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	•	•		1
A	64:4 168:18	127:15,19	accumulating	282:6 301:12
abandon 19:10	absolutely 42:18	130:6 131:7	33:4 139:17	336:12
38:13	84:14 132:21	132:4,18	accumulation	acting 107:11
abandoning	146:16,19	146:10 153:16	45:7	208:22
288:5	151:4 152:3	162:13 168:6	accuracy 93:14	action 28:1
abide 158:16	177:21 194:17	170:12 188:20	93:17,19	69:19 281:21
172:21	254:22 284:2	191:5 195:21	accurate 41:9	298:12 385:11
ability 44:17	287:15 313:17	221:12,15	48:15 84:14	actions 22:11,19
48:12 111:2	316:12 327:3	233:8 241:6,7	93:9,12 94:1	198:17
121:1 161:3	327:18 341:7	249:15 293:5	158:21 159:12	active 135:16
165:9 195:13	348:13 350:19	300:10 306:4	accurately	284:15
228:20 239:15	353:1 374:7	307:17,18,21	18:22 99:3	actively 113:5
266:4 340:12	375:16 378:9	307:22 308:1,1	achieve 86:4	144:3 145:20
377:3 380:13	absolutist 73:8	308:4 310:13	114:17 289:2	161:16
able 23:11 41:11	abundance	341:10 342:18	313:22 314:3	activities 14:2
64:5 89:14	139:12	349:20 376:22	316:16,19	28:20 39:13
90:19 135:12	abundant	accessed 94:5	317:11,12	45:15 46:14,21
146:2,2,5	138:14	190:20 306:22	323:9 335:18	50:6 128:13
155:5,6 166:8	abuse 30:18,19	accesses 39:9	345:7	130:8 182:12
166:10 170:11	31:7 62:6 85:4	168:1	achieved 345:12	183:5 208:4,5
170:19 181:8	86:6 94:18	accessible 75:1	achieving	210:2,11
182:8,10 183:3	113:22 169:18	162:5	304:13	211:13,16
184:9 188:22	abused 55:16	accomplish	acknowledged	214:5,20
190:1 195:6,21	abuser 114:2	291:16 300:16	95:16 123:10	217:10 218:2
202:18 205:17	academeia	300:21	132:8	219:13,21
208:11,15	234:13	account 36:13	acknowledge	223:8 224:14
221:8 270:16	academics 274:7	37:2,5 68:16	84:11,15	224:15 225:18
270:18 279:15	382:21 383:9	104:6 225:3	acknowledges	226:13 230:8
285:21 304:14	academy 181:15	231:7 247:8	292:9	231:6 242:14
307:1,2 309:22	313:10	268:20 318:1	acquire 42:20	246:14 247:5
310:9 311:22	accept 40:6 70:6	325:11	345:1,16	251:8 254:19
312:2 317:7	98:16 99:9	accountability	acquired 151:1	268:17 271:19
338:22 348:19	101:4 375:11	21:19 22:1	acquiring 323:6	272:1 279:15
348:21	acceptance	24:13,17 52:16	acquisition	302:8 383:22
ably 101:21,21	213:7	71:13 72:19	43:12 323:6	384:8
abroad 227:17	accepted 95:8	73:18,19 83:22	acromyn 220:2	activity 5:1
absence 173:21	97:5 106:11	84:8 85:7 86:2	acronyms 296:5	20:18 21:3,10
346:5	149:19 204:8	94:6 126:5	act 14:7 21:7	31:2 39:11
absent 335:14	accepting 350:7	127:9 129:2	22:6,7,18	53:17 65:3
335:14 345:10	350:8 384:5	223:1 225:4	77:22 103:6,14	70:22 183:13
347:9 362:4	access 45:4 75:9	276:17 345:18	130:14 212:6	212:5 218:11
absolute 23:10	78:11 93:2	346:17 383:14	216:5 221:10	233:1,4 245:2
31:22 32:9	104:19 112:15	accountable	222:12 229:19	248:18 260:22
	115:6 127:14	237:2 346:14	240:21 281:10	262:4 265:7,11

201 17	220 12 221 12	1. 4. 10.1	202.10	201 17 271 5
281:17	230:12 231:12	adopts 12:1	383:10	281:17 371:5
actors 94:21	236:1 240:6,7	advance 10:10	affairs 33:12	371:18 372:1
107:15 228:7	270:4 272:14	69:14,16 86:12	189:17	agents 107:12
308:20	297:11 382:17	advanced 40:14	affect 20:1	370:14,21
acts 26:20	addressed 3:10	91:17 170:8	202:9 212:13	371:12
174:18	6:5 20:22	advances 27:5	312:2 338:7	aggdom 217:20
actual 154:7	179:18 203:22	226:15 228:21	affiliation 37:13	220:4
156:4 291:19	230:14	advancing 77:5	afraid 336:2	aggregate 61:3
322:12 332:5	addresses 49:20	77:6 187:16	afternoon 203:4	61:7 77:22
360:6	adequate 24:12	234:4	203:7 284:22	89:7,15 163:10
adage 247:12	24:17 29:22	advantage 12:18	afterward 38:21	164:12 258:9
adapt 146:6	30:11 61:17	317:17 382:1	age 25:15 74:21	347:2
149:17	83:21 154:19	382:18	168:7 186:6	aggregated
adapting 277:22	adequately	advantages	297:9 340:6	77:15 129:17
278:1	22:22 72:18	38:18	362:20 370:17	236:13 323:10
add 10:20	94:7 211:4	adversarial	agencies 39:3	aggregating
119:10 198:11	adequatly 262:2	130:13 212:10	40:13 41:4	40:2 256:8
244:11 296:12	adhering 125:15	adversaries 89:6	74:2 81:9 91:1	aggregation
315:15 329:8	adjourn 384:10	208:14,16	114:10 133:1,5	47:4 48:2
344:12 361:19	adjourned	267:11	205:9 208:14	50:17 61:15
380:12	384:13,15	adversary	208:22 211:10	77:3,14 78:1
added 307:7	adjourning	135:16 280:7,7	212:14,17	161:3 258:7
adding 60:12	384:11	373:11 376:9	241:7,17	335:6
354:19 373:10	adjudicated	380:10	244:19 248:20	ago 123:14
addition 115:21	70:16	adverse 56:3	264:20 267:8	137:16 149:19
239:22 298:21	adjudicator	60:7,18 62:4	298:5 299:17	167:10 229:9
358:17 361:7	29:18	adversity 55:9	306:11 319:14	288:6 306:12
additional 12:18	adjust 100:14	advertised	358:22 359:10	347:15
81:9 154:13	admendment	295:11	agency 3:14	agree 8:9 31:13
181:9 185:7,8	281:20	advice 172:3	63:3 189:16	50:11 52:2
236:18 258:9	administration	207:13 216:3	209:10,14	65:22 77:8
270:19,20	44:15,20	225:15 241:15	217:7 222:13	78:3,13 84:17
274:10 331:20	128:18 222:6	advising 124:19	226:1 242:3	87:10 89:16
333:1 370:11	282:7	advisor 25:6	249:15 251:18	96:20 97:1
additionally	administrations	215:13	256:16 264:5	98:3,10 108:2
129:9 228:19	360:19	advocacy 42:6	273:22 275:11	125:21 146:13
addon 198:19	administrative	295:15 376:1	277:11,12	149:13 186:16
address 26:5	29:21 30:9,12	advocate 85:21	319:8 335:12	189:9,20
46:20 47:1,21	59:3 95:7	212:9 240:16	353:7 354:1,21	281:22 296:15
50:7 67:5	administrativ	244:18 358:18	355:13 374:14	313:17 314:6
105:4,7 114:13	281:7 282:10	359:1 376:8,13	382:22	315:9,10
132:8 163:9	adopt 145:9	376:18 377:17	agencys 221:13	316:11 337:9
164:11 182:1	adopted 100:4	advocated 168:4	225:12	350:20 375:16
207:8 217:13	adoption 125:8	advocates 274:7	agent 23:6	380:2
	l ⁻	<u> </u>	l	l

agreed 6:19	226:15	383:12	308:13 310:19	132:7 157:1
331:9,10,14	alphabetical	amendments	312:19 313:1	anonymity 5:3
agreement 97:4	109:18	87:22 314:12	321:18 323:3	anonymization
97:19 100:8	alteration	american 16:12	324:11 325:5	113:6,13,19
129:16	337:21	131:6 163:14	330:21 349:7	159:4 162:7
ahead 51:2	alternate 30:12	168:2 197:15	350:17 367:11	175:16 196:9
201:21 202:2	298:4	214:6 252:19	370:13 372:17	anonymized
331:18 353:16	alternative	278:18 352:3	382:5	113:10 196:17
aim 21:15	381:11	americans	analyst 174:6	anonymous
air 324:17	alternatives	119:22 122:7	323:11 325:15	198:12 269:16
airbags 40:8	113:3	122:16 200:22	analysts 256:11	answer 26:9,10
airplanes	alters 337:12	201:3 363:13	321:17 370:14	58:10 76:10
294:19	altogether 32:7	amici 376:17	370:21	102:18 103:9
airport 58:15	alvaro 3:4 117:3	377:1 378:7	analytic 39:4	107:8 155:14
akin 26:16	117:3,8 146:21	amicus 375:3	304:13 320:10	175:21 197:9
alex 3:11 203:14	166:18	376:16	323:12	199:15 237:18
206:4,16,22	amassing	amount 23:1	analytics 35:5	238:16 267:14
219:16 223:20	104:17	63:7 72:18	35:17 47:9	327:9 329:1
241:4 245:8	amazing 206:9	91:4 137:18	114:22 304:5	338:9,12 344:2
246:22 247:11	amazon 93:13	150:2 152:12	305:20	358:16 367:14
252:17 262:5	93:15	178:4 182:16	analyze 34:3	368:8 370:9
264:6 275:21	ambitious 7:22	306:8 320:6	105:1 117:16	371:1 372:5
278:16 310:5	amend 282:6	371:7,13	228:20 232:2	answering
alike 137:8	amended 221:15	amounts 186:17	320:2,3	141:15 230:5
alleged 130:20	amendment	310:7,15 320:3	analyzed 37:11	318:15
241:9	12:20 14:22	320:7	366:20	answers 76:2
allies 227:15	27:22 29:13	analogy 40:4	analyzing 40:3	108:11 313:16
allow 91:17	42:2,14 43:6	133:19 194:19	233:14 312:20	anticipate 58:11
292:10 298:21	52:10 53:7,14	346:4	366:13 368:9	181:17 282:18
298:22 301:16	64:13,22 70:1	analysis 11:7	ancient 8:10	antilock 40:8
350:1	71:15,17,22	37:17 40:18,22	anecdotal 311:4	antiques 25:22
allowed 78:21	72:6,7,8,21	41:9 50:22	anecdote 310:17	antiterrorism
243:20 355:5	80:19 81:8	62:9,12 65:17	anecdotes	182:21
364:8 369:3	82:1,2 103:18	90:19 91:7,13	312:20	anton 3:3
373:8	107:5,9 121:5	91:18 94:3	anew 53:12	110:12,19
allows 14:8	121:6 167:14	101:13 105:13	angeles 182:7	116:17,22
34:21 38:20	167:19 168:19	111:12 115:13	animate 98:14	142:8,21
52:10,10	221:12 229:16	117:15 229:17	animates 85:4	144:11,18
132:17 277:19	292:6,13	229:20 230:4	annie 3:3 110:12	145:14,17
307:14	293:14,18	231:2 233:11	110:18 165:18	146:16,19
alluded 172:6	294:3,5 314:15	233:20 236:18	176:1	159:2 160:1,13
185:12	334:18 350:12	260:18 263:8	announce	162:1 164:16
ally 376:14	353:12,14	269:22 270:19	161:16	166:5 175:17
alongside	368:22 372:16	273:20 293:22	announced 4:9	176:17 177:21
	I		I	I

				1
184:7 185:3	107:10 135:2	approaching	argue 63:3,9	87:2 100:12,20
186:22 187:1	272:1	211:6 295:7	89:10 118:19	100:22 105:20
194:7,11,17	apply 46:1	appropriate	317:9 349:3	182:14 196:15
195:19 198:11	64:11 75:16	25:13 32:19	argued 38:12	229:21 233:3
199:14 200:13	80:22 81:11,14	71:12 73:1	argues 150:13	238:19 262:11
antons 189:10	82:9,13,16	94:5 168:13	arguing 129:11	282:20 288:20
anybody 55:16	86:18 113:7	169:6 258:20	argument 18:11	289:4 318:18
58:7 64:9 81:4	114:3 119:15	261:3 374:15	18:12 43:14	343:16 363:5
164:14 182:1	162:17 163:3,4	appropriately	90:2 120:8	asks 41:2 104:5
244:1 339:13	187:14 191:10	52:22 73:7	167:7 310:18	aspect 48:10
anymore 149:21	196:10 202:14	94:1 130:6	311:4 369:19	58:4 236:15
209:20	208:2 220:1	209:1 248:6	arguments 18:9	345:22 365:2
anyway 64:21	260:7 271:9,10	349:1	18:12 268:5	aspects 8:9 11:5
106:16 333:14	271:18,20	appropriation	arises 29:1	15:19 42:1,5
apart 256:4	272:4,8 273:1	5:9	arm 56:20	45:16 54:12
apologies 33:10	276:1,10	approval 219:11	arms 68:18	55:2 59:11
333:14	297:16 299:16	222:8 223:11	298:18	101:12 199:18
apologize 10:10	320:1,2 327:3	281:13 308:15	arrangements	211:15 248:13
app 350:15	343:18	308:18	325:20	249:3 251:19
apparatus 89:14	applying 3:17	approved 222:7	array 62:15	315:11
apparent 199:6	46:11 82:7,10	314:22	319:4	aspersions
266:8 273:22	112:12 271:15	arbitrary 102:5	art 233:17,19	123:16
apparently	300:3 327:19	architect 3:7	234:17 236:16	assembled
352:11,13	appreciate	62:20 134:15	261:14 270:14	204:22
appeals 364:14	25:11 116:3	architecture	316:8 335:14	assert 112:11
appear 6:19	132:6,21	217:5 221:1	article 45:12	asserting 191:20
338:19	134:21 165:16	225:5	46:6	assertion 27:4
appears 17:15	215:8	archive 306:16	articles 54:3	assess 205:5
80:8	approach 5:18	archives 222:6	195:13	231:18 236:9
applaud 287:16	12:8 40:1	area 96:5	articulable	254:19
apple 112:3	53:21,22 61:21	105:20 184:8	60:13	assessesment
applicable	72:7 88:3	184:18 200:7	articulated	218:20
182:20 216:4	98:20 105:8	280:9,18	272:3 331:17	assessing 245:12
223:21	111:11 125:3	293:11 327:20	articulating	245:15
application	126:17,20	336:5 368:11	204:3	assessment
106:18 299:20	173:14,15	areas 46:14	arts 235:12	221:20 223:13
303:3 332:2	187:8 231:2,4	69:19 192:9	ascertain 363:10	231:14 232:20
365:13	244:3 289:16	276:5 294:9	363:21	233:6 261:12
applications	291:13 329:21	arent 20:17	aside 58:17	290:5,18
32:17 137:2	330:4,5 370:19	24:15 29:19	97:21 317:15	331:16 332:5
143:5 159:5	382:2	166:16 173:4	asked 157:9	370:15
applied 6:7 72:1	approaches 39:3	200:15 309:15	216:10 219:22	assessments
333:8,8 369:13	172:8,9 235:7	321:4 331:6	239:3 363:5	204:10 219:13
applies 42:15	331:6	354:22	asking 80:18	220:13 223:9
<u> </u>				

224:21 240:20	attempted 282:5	246:13 276:18	160:19 161:10	184:12 203:9
241:1 325:21	attempting	308:22 321:1	173:8 185:2	217:11 255:13
assist 254:2	11:14 174:7,11	321:22	186:9,18 187:2	284:15 286:15
325:13	attempts 68:20	auditors 324:21	188:18 189:3,4	305:12 318:9
assistant 25:7	131:8 212:4	325:6	189:5 205:18	318:16 321:3
136:10	255:17,17,18	audits 56:8	219:5 220:13	334:9 341:16
associate 124:22	attendance	372:22	250:2 287:1	347:13 354:14
associated 194:9	16:19	authenticated	295:4 306:11	356:14 366:6
214:4 232:22	attends 192:8	112:16	325:17 339:1	366:12 373:20
247:19 324:8	attention 5:4	authorities	339:13	374:11
association 11:3	95:11 150:2	204:17,18	avalanche 34:18	backdoor 166:1
associations	155:15 157:4,5	209:17 229:18	35:13 37:2	166:9 187:9,16
16:19,19 153:3	158:6 243:8	302:6,9 323:5	ave 1:13	300:12 314:20
assume 64:11	286:6 287:12	364:14 365:5,7	avenues 356:4	backdoors
107:13 333:21	292:2 293:8	authority	average 141:8	111:18 112:17
assuming 92:19	344:18 362:9	107:11 219:20	152:14	backed 270:22
256:20	382:6,20	242:13 301:20	avoid 68:8	backend 300:4
assumption	attitudes 340:3	314:17 322:18	208:16 366:4	359:4 372:4
186:17 254:5	attorney 29:5,8	324:4 345:16	avoiding 5:6	background
assurances	112:10 206:17	362:22 364:16	avoids 12:14	36:5 106:22
130:17 131:13	215:13,17	authorization	awang 7:15	135:6
212:17	217:19	351:9 352:4	aware 79:17,18	backwards 60:1
assure 55:15	attorneys	authorize	153:12,17,18	bad 18:8,15
182:14,15	237:12	302:12	154:14 157:12	20:18 24:15
astonishing	attracts 95:11	authorized	173:8 252:20	46:22 161:6
330:13 351:16	attribute 346:9	112:16 131:7	266:12	174:18 188:10
astray 335:15	attributes	265:1 334:19	awareness 79:15	201:19 263:4
atheist 140:18	323:19 324:2	authors 16:13	172:10 174:21	275:1 301:18
atmosphere	audience 6:13	automatically	309:19	310:21 315:4
54:18	9:21 88:9	142:6 245:4	awful 43:11	360:15
atomic 326:2	102:3 109:1,9	automating	96:11	bag 307:9
atomically	110:5,7 141:22	308:14	axes 45:3	bailiwick 86:10
324:9 330:15	154:12 181:12	automotive 26:4	aye 4:18 384:11	106:15
attach 34:10	278:7,11,15	autonomously		balance 8:17
attack 141:8	281:3 286:12	142:6	B	19:20 22:22
161:16 170:13	318:8,13 383:5	autonomy 27:4	b 91:10 198:4	33:5 52:4 53:1
170:20 195:5	384:3	availability	back 8:10 25:20	53:7,11,13
attackers 112:1	audio 329:16	194:15	57:12 76:21	64:5 73:6,9,10
attacks 111:16	audit 31:19 93:3	available 14:15	80:17 87:15,17	73:10,12 133:8
112:15 138:1	94:4 95:3	14:19 15:3	87:19,22	133:12 134:1,4
141:3,6,10,12	308:9 310:12	32:22 35:19	106:21 134:12	134:6 167:20
169:14 170:4,7	320:7,17,18	36:21 40:7	138:5 142:9	169:20 170:3
attempt 7:21	321:1,6 348:9	75:5,12 90:8	156:1 163:12	176:14 178:18
130:17 212:12	auditing 57:21	92:6 106:4,13	171:12 181:8	260:4,6 302:21
	ı 		1	<u> </u>

		_		
319:16 337:4	basis 54:6	behave 37:21	best 12:10,11	269:7 288:6
337:12 345:8	184:20 209:16	95:20	38:6 63:15	289:13 297:9
345:12 383:19	233:13 235:1	behavior 16:7	88:3 105:8	317:1,3 319:22
balanced 19:18	240:17 274:18	16:10 34:20	112:12,22	321:22 374:3
130:7	299:3 307:19	35:1,3 68:7,8	148:2 177:20	bigger 257:5
balancing 19:15	317:8 364:6	68:19 69:3	197:22 213:7	biggest 92:8
25:2 52:14	bates 373:15	85:6 280:13	213:10 251:17	121:8 241:15
53:22 64:12,14	377:11	behavioral 36:2	258:16 296:5	267:22
64:22 65:16,18	battle 243:6	behemoths 44:1	309:10 343:8	bilaterally
65:21 66:13	bay 113:18	belief 125:4	344:3 373:20	153:14
67:16 68:15	beach 300:11	beliefs 153:4	380:18	bill 8:12
72:9 204:4	beaker 347:10	believe 17:6	beta 252:12	billion 141:9
291:9	bear 239:8	32:8 65:22	beth 88:8 203:8	binary 198:19
ballot 27:17	244:16 314:8	113:12 115:15	219:22 278:11	bind 323:18
ballroom 4:6	bearing 378:10	127:13 130:1	better 25:2	324:2 326:2
baltimore 385:5	bears 89:4	131:4 137:17	31:10 50:3	binding 330:15
ban 123:12	becky 225:11	150:20 151:2	53:1 112:14	331:6
180:12	237:22 249:21	164:2 179:14	129:21 143:15	biographies
banning 121:15	260:2 261:15	184:14 187:1	146:6,14 166:5	286:22
bar 29:16	262:8 268:9	230:22 235:5	182:15 187:9	biological
base 167:15	276:7 282:15	254:11 313:18	189:11,17,19	227:12
based 9:7 17:18	354:19	317:1 330:9	191:12 195:19	biometry 170:14
18:13 44:10	beckys 255:16	345:12 368:20	204:15 208:15	bios 10:3
48:19 53:17	becoming 13:16	372:20,21	235:3 237:7	bit 45:6,14
65:13 81:21	138:21,22	believes 42:6	258:12,15	53:21 54:19
125:20 147:5	139:3 147:18	296:8 297:13	264:3 267:5,6	57:18 92:18
156:11 164:4	148:9 170:8	beltway 197:4	292:16 358:14	159:3 162:19
176:5 206:15	182:15,17	benchmark	373:22	165:19 179:20
230:7 237:4	bedoya 3:4	119:7	beyond 21:14	190:21 204:7
252:3 363:12	117:3,9 146:22	bend 256:17	39:22 61:20	205:20 213:4
366:5	147:14 162:15	beneficial	74:22 82:1	217:9 218:18
baseline 81:8	162:20 167:5	274:16 381:18	312:8,19 335:9	233:16,18
218:12	178:9,19	benefit 50:22	366:4 367:13	246:10 248:10
basement	183:19 187:20	123:5 202:12	bias 313:1	252:2 262:15
306:16	192:20 196:18	286:2 310:18	biases 315:14	271:7 277:4
basic 61:8	197:8 200:20	318:13 330:21	bieber 368:3	288:8 305:14
136:18 205:22	201:2	336:14	big 35:4 44:1,22	347:5,14
234:22 256:18	beginning	benefits 50:13	49:16 50:14	bits 268:14
257:14 283:4	171:22 205:11	50:21 52:2,5	52:2,8,14,21	black 307:12
342:15,16	233:21 273:11	52:14,18,21	77:18,20 93:16	blank 334:12
basically 72:9	274:17 275:14	106:22 107:2,3	114:22 186:5	blend 196:6
104:22 121:17	337:3	112:21 277:12	190:11 228:20	234:17
138:18 166:11	begins 60:17	291:16 331:17	233:22 234:20	blending 233:16
200:6 273:16	behalf 65:21	benign 152:10	237:8 257:4	blind 194:3
200.0 273.10		~~gii 132.10	257.0 257.4	71110 17 1.3

361:20	73:3 74:7	331:18 333:11	broadening	135:3 138:16
blinking 22:14	boil 232:8	335:22 338:10	253:7	140:4 155:8
block 183:12	boiled 347:12	340:7 341:16	broader 33:6	157:20 158:3
blocks 235:10	bolton 346:7	344:10 347:16	231:2,12	232:16 235:10
236:14	bolts 96:1	363:2 372:8	289:17	236:14 237:4
blurring 108:5	book 93:16	381:6 383:1	broadly 19:9	239:5 251:21
board 1:3 2:1	288:7 306:20	breach 62:6	45:17 240:21	305:1 312:1
4:12,13 6:10	books 93:13	169:19	broke 16:17	347:22 357:5,8
6:10,11 9:18	boost 302:1	breached 300:7	brought 57:15	builds 134:16
10:12 11:13	border 132:18	breaches 38:3	238:13 341:17	304:4
25:11,14 26:14	290:5	breadth 245:9	brown 3:13	built 105:14
27:15 28:3	boss 23:5 193:15	251:9 278:21	215:10,17	126:9,11
32:5,15 41:2	bottom 50:4	break 7:4 109:2	216:7 239:17	136:17 157:18
51:12 56:9,13	273:16 326:1	187:14 195:3	245:22 263:11	201:16 214:12
74:12 85:20	bound 57:16	284:18 346:10	267:13 277:16	214:13 230:6
96:14 98:15	313:9 324:9	brennan 2:13	281:9	242:8 362:12
101:6 110:4	345:18	10:7	brute 61:21	bulk 15:15
116:10 120:21	bounds 22:8	brief 7:2 18:6	89:21 90:3	90:17 114:13
128:15 140:12	42:8	31:12 50:9	195:5	128:7,9,9,12
141:17,19	box 192:12,19	76:2 79:6	bubbles 245:2	180:2,5,12,18
159:15 163:12	277:2,5 320:17	102:7 165:17	bubbling 263:3	180:21 200:21
163:21 176:3,4	boxes 232:21	175:21	build 35:21	201:3 383:15
181:9 185:7,7	bradfordfran	briefings 193:12	37:22 41:4,11	bullet 172:7
192:1 207:2,8	7:14	253:20 374:12	56:7 61:2 85:8	332:9 370:8
207:10,17	brakes 40:8	briefly 10:5 53:4	112:3 125:14	burden 322:18
209:20 213:18	branch 2:14	87:8 92:17	127:12 135:11	323:4 330:2
213:19 216:8	25:5 212:22	200:7 299:15	135:17 138:11	burdens 324:6
236:20 252:16	223:4 227:9	305:18	138:18 140:14	burdensome
282:15 286:10	246:20	briefs 375:3	142:1 155:1,7	74:20
287:16 295:21	branches 73:20	379:7	155:11,12	bureau 240:9
303:20 304:1	315:2 360:16	bring 88:10	156:2,7,11	bureaucracy
357:3 367:17	brand 2:4 4:14	110:9 216:18	158:10 170:22	351:21,22
383:21,22	80:14,15 81:22	231:1 239:8	174:20,22	352:1 365:9
boarding	82:21 83:3	244:13,14,16	199:17 202:7	bureaucrat
294:18	87:6 159:18,19	314:8 318:2	236:17,22	22:11
boards 4:4 5:18	161:18 162:15	327:7 378:6	247:10 271:2	bureaucratic
7:13 32:20	164:14 165:13	bringing 165:11	307:18 311:19	353:22
203:5 207:11	192:3 271:5	376:17,17	346:7 348:3	burgeoning
207:18 287:4	272:10 275:20	brings 50:20,21	349:4 356:4	78:18
349:13	276:12 285:7,9	broad 12:2	357:13	burglars 194:21
bodes 88:16	295:13 303:8	61:16 152:7	buildable 135:4	195:2
bodies 32:21	313:3 318:7	240:22 251:9	140:21 158:5	buried 27:7
72:5 94:11	322:13 323:21	262:18 299:22	171:4	burn 307:9
body 33:2 49:1	326:7 328:2	324:15 383:11	building 86:12	business 11:14
	ı	ı	1	1

108:13 125:17	cant 17:6 22:22	286:4,17	cat 8:5	375:3
126:10 127:2	23:6,16 32:1	287:10 318:9	catagorization	cell 194:12
264:22 265:17	32:10 48:20	care 8:15 15:20	236:6	cellphones
265:22 266:4	53:8 59:19	15:22 17:8	catagorize	329:15
266:14	64:4 65:9	138:15 149:20	235:14,16	census 122:9,15
businesses	70:20 71:4	150:1 197:1,4	236:4	123:10,11
265:3	76:10 77:16	197:5 198:20	catagorizing	center 2:13 3:4
button 201:22	79:15 84:16	199:7 247:1	268:10	3:19 33:13
349:5	86:22 94:8	355:3 356:20	catalogue 27:9	40:3 62:2
buy 24:7 25:1	107:14 116:1	career 247:20	catch 173:15	117:4 132:13
buyin 23:20,21	145:4 162:18	careful 207:19	175:6 311:22	156:3 232:5
buying 348:16	163:3 174:17	210:12 248:10	catches 311:21	295:16 296:7
	188:21 198:20	294:1	catching 190:6	centers 10:7
<u>C</u>	241:17 261:22	carefully 88:1	cate 3:18 286:21	130:18,21
c 1:14 4:8	262:1 271:10	140:11	287:2,6,8	186:1 188:4
124:22 248:1	271:18 272:4,8	cares 292:19	296:14 326:9	centralized
cabinets 151:10	278:18 288:19	carnegie 251:8	328:2,16	127:1
cables 130:18,20	311:4 326:20	carried 136:15	335:22 336:2	centric 234:3
cake 86:15	330:9 331:9	carry 137:10	341:17 350:19	century 337:17
calibrated	349:2,4 367:11	214:19 259:2	351:12 354:18	ceo 357:12,14
218:15	capabilities	265:6	355:8 374:4,5	certain 13:7
california	170:6 171:7	carrying 383:18	375:17 381:9	16:15 37:5,6
133:14	191:20 309:5	cars 106:1	381:12	42:1 46:14
call 4:16 37:10	309:15 348:15	case 65:9 67:4	categories 152:2	52:12 63:4
48:10 49:1	357:5	76:8 81:17	306:2	68:9,9 71:5
54:2 85:14	capability	86:11 95:21	cates 305:22	73:17 76:15
121:18 139:2	279:14 345:1	133:14 166:7	cato 66:16	83:12 84:4
152:11 156:14	capablilities	168:11 184:14	cats 368:3	85:18 102:21
156:21 175:11	234:10	218:16 259:19	caught 310:20	103:15 106:6
180:18 289:13	capacity 41:5,12	264:7 270:15	cause 20:15,20	159:5 160:14
289:14 382:9	206:10 215:12	281:4 291:22	29:17 46:14,15	165:8 173:3
called 45:13	capital 3:20	295:5 326:21	46:22 47:1	174:3 184:13
54:7 120:22	313:5	331:10	53:17 65:3,14	190:2 220:10
132:15 156:15	car 26:2 106:3	cases 48:12	73:1 81:21	224:4 235:21
198:13 253:10	321:8	53:15 71:19	104:18 169:5	268:10 279:7
288:7 317:2 335:20	card 6:15 9:12	83:7,12 113:20	362:3	380:22
	9:12 17:21	114:7 153:10	caused 50:6	certainly 17:9
calls 32:3	37:5 59:22	160:14,22	104:15	17:15 43:2
163:20 377:4 camera 170:17	66:14 67:10	162:10 171:3	causes 16:6	45:2,4 58:5
camera 170:17 cameras 170:15	110:1,2 206:7	191:4 265:9	20:19 47:17	59:9 60:7 64:3
318:14 369:17	286:5,6,17	270:9 376:10	237:3 359:13	70:21 84:17
	293:17 354:4	377:7,14	causing 17:3	94:14 98:16
camps 122:17	cards 6:15 110:7	cast 52:4 123:16	cdt 297:13	108:14 146:13
candor 29:7	229:11 255:3	castles 300:10	299:11 303:1	149:5,15,17

				1
153:19 154:9	challenges 26:1	cheaper 138:13	77:22	229:14 231:1,3
157:15 158:7	226:19 241:16	148:10 169:13	chosen 13:9	231:12,14,18
168:3,12	279:12,17,22	185:16,17	chris 3:20 313:4	231:22 232:4
178:19 179:16	327:19 378:20	check 59:5	348:1 351:6,15	232:17,19
180:22 184:7	challenging	277:2,5 325:3	358:1,13	233:4,10 234:5
188:2 192:20	204:6 226:17	checked 277:14	360:11 364:1	234:16 237:6,7
197:11 198:5	championed	checking 192:12	church 73:22	237:12 238:10
224:7 230:12	50:16	192:19 232:21	210:9	238:12 240:3
243:17,22	chance 13:12,13	checkpoints	circling 262:14	242:15,21
254:20 276:2	333:16	126:13 172:22	circulate 9:22	243:8,19
282:5 288:15	change 16:7	173:3	circumstance	244:10,13,18
289:9 293:15	123:9 137:20	checks 173:10	12:5	275:7 280:22
296:13,15	145:18,21	175:2 320:17	circumstances	282:21 284:13
304:21 317:20	146:1 193:1	chemical 227:12	13:6 64:20	289:19 295:17
333:7 341:5	280:13 288:6	chertoff 25:6	80:1 84:17	295:21 296:2
355:17 356:14	315:21,22	cheshire 8:5	99:14 101:8	303:9 304:18
377:13	322:11 341:8	chew 383:16	circumvent	305:5 310:21
certainty 159:13	341:11 361:3	chicago 183:9	175:5	317:10,12
certification	381:1,1	183:10	cited 375:3	334:1 349:13
376:6 385:1	changed 16:10	chief 3:7 117:6	citizen 140:10	351:18 353:8
certify 385:6,9	17:15 70:13	124:11 133:13	201:21 337:18	356:21 357:6
chain 189:3	137:5,6,8	134:15 215:10	337:20	358:21 372:17
266:11 308:15	149:15 190:4,5	223:17 240:4	citizenry 314:8	383:19
349:6,7	227:22 228:1	childhood 13:8	337:6	civilizations
chair 74:13	228:10 229:5	13:13	citizens 132:9	8:11
108:17 110:13	334:5 341:11	children 303:22	133:10 167:17	claims 9:7
chairman 2:3	changes 137:13	childrens 341:5	302:20	clarify 180:17
4:11 25:10	147:8,11 215:3	chill 21:3,4,5	citizenship	clash 20:22
27:10 101:21	226:14 278:1	chilling 21:9	230:21 232:5	class 136:3
109:8 165:14	364:7	79:2	civial 215:14	classic 329:4
201:8 383:2	changing 33:19	china 170:18	civil 1:3 4:3	classified
challange 253:2	110:21 277:18	chinese 189:13	111:4 116:9	253:19 353:4
challenge 83:16	348:20	choice 23:13,16	135:15 203:5	356:8
84:21 92:9	channels 223:16	23:17,18 66:1	205:10 206:5	clause 64:14
200:18 221:6	228:15 280:12	66:2,3,4 78:2	206:11 207:16	clear 90:20
222:10 241:20	chaplains	126:3 255:9	209:2,13,19	123:8,15 128:8
241:22 264:20	122:21	256:22 289:11	211:3 215:11	131:12 133:16
265:11 267:22	chapter 288:7	299:9 307:21	215:18,20	149:22 158:9
328:19,21	characteristics	choices 106:10	216:1 217:16	158:10 191:5
335:18 353:9	46:10	199:4 270:3	222:21 223:17	202:2 230:11
374:18 379:9	characterize	choose 297:2	223:19 225:12	295:2 331:15
380:1	209:8	299:5 319:7	225:16,18	337:21 370:9
challenged	charged 360:21	326:20	226:6,9 227:2	375:10 378:8
132:11 329:11	chatty 139:4	chooses 77:22	227:18 229:4	381:12 382:19
	•	•	•	•

				1
clearance 193:9	cold 10:10 228:2	200:2 208:15	296:17,20	287:9
201:7 355:2	collaboratively	281:18 283:9	297:7,14,15,21	com 93:13
clearances	246:21	301:14 339:15	299:1,7,8,11	comb 322:2
193:6	colleagues	350:16 353:18	299:21 300:8	combat 313:21
cleared 85:21	148:20 216:21	collection 15:15	300:15,22	combination
253:3,13,16	collect 9:22 34:2	15:15 20:13,17	301:7,8,16	268:15
254:2,9,11,14	40:15 61:22	30:20,22 34:5	302:18 303:6	combinations
254:18,21	62:4,11,20	36:11 37:18	305:21 317:2,4	268:20 269:1
clearer 146:14	63:4 70:2	38:13,17,19,20	323:2 324:11	combinatorial
291:3	79:22 87:7	38:22 39:2	325:5 327:15	138:3
clearing 254:5	106:9 118:1,14	40:19 46:4	333:9 338:3	combine 36:15
clearly 80:10	118:15 120:9	56:2 61:15	350:11,12,17	47:7 75:13
119:15 163:2	121:14 178:21	63:10 66:9	350:21,22	76:12 335:1
219:2 223:12	179:1,3,10	69:8 77:14,17	351:3,8 359:17	combined 41:8
249:9 331:16	277:19 292:22	77:20 78:13,22	360:3,6 361:14	217:22
332:17 336:7	293:4 298:3	82:6 84:7,16	371:9 372:6,10	combing 14:16
375:6	324:1 342:2	87:12 89:12,22	372:20 383:12	combining
clie 136:9	352:22 353:1	90:4,9,11	collections	76:13
client 29:5,8	359:21 360:7	92:21 102:13	180:2	come 16:22
354:10	382:12,14	102:14,15	collectively 11:4	29:11 57:15,17
clients 267:18	collected 5:11	104:17 108:4,4	collectors 237:1	77:6 104:12
268:2	15:12 19:8	108:7 118:7,20	collects 15:21	124:18 132:16
close 13:8	21:15 37:11	118:21 120:4	16:3 298:11	134:12 140:20
140:22 141:14	63:7 79:18	120:20 124:4,7	300:8 317:19	142:9 180:9,11
closely 88:4	84:3 90:14	126:3 128:7,10	college 137:2	180:14 216:12
133:2 207:11	99:19,19 118:9	128:12 136:6	157:9,10	217:16 228:7
closer 323:22	121:16 123:3	147:19 169:12	collins 2:7 4:14	229:5 257:14
closing 115:14	126:2 127:16	169:12,21	88:14 92:8	259:21 271:22
cloud 125:8	152:8 163:16	178:4,4,10,14	171:10 173:17	272:21 289:10
136:22 151:11	164:5 169:11	178:14,14,17	175:8,18 196:8	305:6 319:11
151:14,17	202:16 204:17	179:7 180:5,12	203:9 206:9	322:19 324:7
188:19	221:19,20	185:18 200:21	207:3 215:9	325:21 347:3
code 187:5,10	229:22 230:20	201:3 228:3	225:8 237:20	367:21 372:18
coded 32:6	232:11 266:13	230:5 232:7	239:10 241:4	373:5,12,18
coding 234:13	278:20 282:20	256:1,3,5,6	241:14 245:7	comes 10:11
codirector 10:7	306:9 324:5	257:21 258:4,6	247:11 249:21	12:9 21:9 25:2
coerce 104:1	330:11 358:2,5	258:21 260:14	252:14 255:1	40:13 53:6
coercive 44:7	366:20 371:6	260:19 262:18	255:10 262:5,7	63:13 123:19
58:22	collecting 36:7	262:20 271:14	263:9 271:4	133:14 149:11
coffee 77:3	36:20 39:4,10	272:6,16	278:2,13	221:2 223:15
coffin 150:12	40:1 55:17	278:22 279:4	280:20 282:14	223:16 245:3
cogent 41:19,20	62:12 75:15	279:20 280:5,9	284:14 363:3	249:16 296:6
cognizable	77:19 118:13	283:7 292:12	366:8 370:10	297:20 323:20
98:14	142:5 165:7	292:13 293:8	colorblind	334:9 345:17
	•	•	•	•

·				1.
	I	I	I	1
358:9 361:4	commission	communities	140:5	239:13,20
comfort 253:9	85:13 216:12	121:12 123:20	compartmented	240:7 241:2
comfortable	216:15 385:18	communitity	253:17	242:21 246:9
24:16,19	commissions	214:12 242:18	compatibility	246:12 316:14
comforted	287:5	community 30:4	222:17	324:21 335:13
177:22	commitment	42:6 90:18	compel 375:9	342:13 343:7
coming 17:21	116:10 125:6	92:2 118:11	compelled 99:15	345:22 346:1,2
88:12 171:6	131:1	146:7 195:4	99:16	compliant 146:3
216:6 243:16	commitments	207:14,15	compensate	complicated
273:2 339:19	126:6	208:8 213:1	100:16 327:13	37:16 63:14
341:1 358:12	committed	214:19 215:5	competing 8:17	171:20 378:2
commencing	282:7	226:9,10,13	9:2	complies 177:14
1:14	committee	242:11 244:2	competition	comply 38:19
commend	30:11 73:22	245:1 254:4	67:12	145:10 178:2
271:13	119:12 163:1	258:3 266:15	competitors	240:14 319:15
comment 50:9	179:17 193:22	266:20 276:11	130:22	342:4
185:14	210:9 303:12	310:22 376:1	compiled 148:8	complying
commentators	committees	companies 24:2	complain 54:4	177:10
8:9	85:13 179:14	44:6,8,21,22	complete 86:2	compoments
commented	194:2 210:7	45:2 83:7	249:13 384:1	224:15
194:8,19	212:20 253:16	90:15 91:3,5	completely	component
comments 5:17	253:21 254:6,9	104:1,4 112:4	31:13 86:6	22:20 48:16
7:9 67:18 76:2	254:16	112:17 127:20	165:1 174:17	76:4 83:20
128:17 134:10	common 18:8	129:9 131:3,6	330:7 337:9	215:20 224:8,8
183:18 186:13	30:3,6 46:9,10	140:1 149:10	350:20	224:9,10
225:7 263:14	115:4 224:15	188:15 298:14	complex 38:9,11	239:21 246:11
313:15 317:15	commonality	298:21 341:22	74:19 94:12	346:12
331:20 383:22	96:12,18	343:14	143:1 171:8	components
384:2,4,5	commonly	company 91:19	308:19 368:5	215:20 224:9
commerce	292:19 293:10	102:21,22	378:20	224:16 246:8
247:22	commonsense	103:2,16	complexity	346:13
commercial	14:21	126:18 134:16	37:19 38:4	comprehensive
39:7,8 43:12	communicating	136:5,12	233:22 245:12	10:18 127:5
75:7,9 108:3	153:1	156:15 172:18	373:17	186:12
111:19 119:4	communication	250:15 304:17	compliance 38:3	comprehensiv
216:16 230:17	189:3 228:14	338:22 350:15	38:5,8,11 95:2	31:2
299:6 304:3	339:11 361:18	356:14 367:2	95:3,16 127:10	compromise
343:6,6 350:14	communicatio	comparatively	143:2 146:11	14:10
commercializ	29:6 43:16	58:18,21	169:15 172:14	compromised
104:10	103:14 128:14	compare 264:9	176:20 177:4	15:8
commercially	166:3 227:7	compared 58:21	192:16 206:19	computer 33:12
39:17	228:12,16,18	comparison	216:4 226:2	33:19 61:11
commerical	249:2 363:16	11:17 67:1	230:6,7,11	74:21 110:15
228:17	370:22	compartment	238:4,8,18	134:17,18
	<u> </u>	<u> </u>	<u> </u>	l

				1
174:10 237:13	concludes	193:7	conscious 357:6	consistentily
295:18 313:10	202:21,22	confidences	consent 79:11	230:3
computers	conclusion	13:18	79:12 80:11	consistently
22:14 269:17	224:1	confidential	81:3,18 98:7,8	95:20 123:18
computing	conclusions	122:12	152:16 221:8	consists 8:19
110:13 234:11	280:10 288:1	confidentiality	288:20 312:14	constantly 54:21
comstat 185:1	347:3 361:5	29:5 122:14	328:6,11 329:5	constituents
conceivable	concrete 92:13	confidentially	331:11 332:13	153:22
256:14	170:9 312:7	123:3	consenting	constitues 235:4
concentrate	concretely 93:4	configure 308:4	103:7 328:9	constituted
45:20 200:10	163:6	configuring	332:19	323:2
concept 8:6,13	concur 293:15	112:5	consents 289:1	constitutes
12:2,11,11,14	concurrences	confinement		319:18
13:3 14:21	76:7	59:6	consequence 55:9 56:3 60:8	constitution
17:5 49:22	conditioned	conflict 67:15		42:1 131:4
			consequences 20:11 36:7	
76:20 80:19,22	364:15	confluence		244:5 313:19
82:9 97:19	conditions 9:6	108:15	60:19 62:4	314:13
120:22 121:4	362:14	confusing 96:16	281:1 334:7	constitutional
138:19 144:15	conduct 31:15	conglomerate	consequential	42:9 98:13
147:5 155:10	73:6 148:10	8:14	54:2	129:12 147:4
196:10 235:11	179:15 194:4	congratulations	consider 5:22	337:3 377:8
366:3	197:3,12	296:4	113:16 114:9	constitutional
conception	198:17 211:12	congress 11:13	177:16 227:18	97:21
18:13,14 72:2	233:4 241:17	12:7 32:22	240:9 247:5	constrained
340:5	363:14	65:20,21 73:22	284:1 344:22	314:9 334:17
conceptions	conducted 70:11	74:4 85:16	383:17	375:20
25:19 48:1	193:12 210:2	95:4 122:14	consideration	constraint 325:3
concepts 61:17	228:13 230:3	123:12 128:19	114:21 143:13	constraints
92:9 171:15	236:19 265:12	130:14 210:4	154:19 176:22	324:6 325:16
conceptualizing	332:6	212:6 230:15	205:11 250:6	345:17,17,19
45:13	conducting	253:5,6,11	253:7 296:15	constructs
concern 61:18	167:1 251:3	254:1,14,16	considerations	324:17
127:15 132:3	321:18 342:21	352:14 378:17	111:7 199:21	consult 176:11
135:3 147:7	conducts 218:19	congressional	231:16 285:6	176:18
188:14 258:12	223:4 225:17	179:14 210:7	303:4 345:4	consulted 242:4
concerned 45:8	227:22	212:20 252:18	379:2	consulting 2:14
56:13 149:9	conference	253:2 303:18	considered	25:5 242:15
188:7,9,11	154:3	321:3	115:7	consumer 366:6
concerns 107:5	conferences	connect 39:13	considering	consumers
127:18 171:15	247:9	connected	231:22 245:9	125:9 298:22
205:10 230:12	confessional	138:21,22	considers 230:4	consumes 208:1
230:13 374:9	27:19	169:14 371:15	364:13	contain 34:16
concluded 45:19	confidence 57:2	371:19	consistent 209:3	34:22 35:2
181:18	57:5 130:5	connecting 49:3	247:6 299:20	contained 385:6
	<u> </u>	<u> </u>	l	l

ir				1
contains 371:14	131:12 134:7	340:18 342:1	247:11 249:21	115:11 133:15
contemplated	181:10 188:16	342:18 349:20	252:14 255:1	141:9 217:16
149:18	198:21 203:6	controlling 13:3	255:10 262:5,7	310:18
content 132:13	211:14 234:3	13:4 86:5	263:9 271:4	costs 68:13,21
151:22 152:19	255:15 280:15	142:15 143:20		342:20
			278:2,13	
153:2 168:6	294:12 355:22	controls 12:20	280:20 282:14	couldnt 81:6,17
188:22 230:19	364:10	38:13 71:13	284:14 363:2,3	104:7 201:1
343:18	continued	93:2 94:8	366:8 370:10	counsel 96:4
contents 36:10	130:10	199:2 209:17	cooks 88:8	117:6 124:12
290:9	continuous	298:17 306:4	cooperation	209:11 238:16
context 3:2 6:1,8	345:22 346:2	307:18 308:2,5	104:20	239:12 243:15
14:6,7 60:6	continuously	310:14 348:2,4	coordination	243:16 244:7
75:17 81:1,12	346:8	348:9,10	91:5	295:15 385:10
82:13 83:11	contract 102:22	353:21,22	copied 290:9	counsels 238:3,7
84:13 85:2	103:1,5 198:10	376:3	copies 361:20,20	240:1
88:21 93:10,20	contractors	controversial	core 29:2 118:22	count 279:14,16
93:22 94:2	104:8	65:6	120:22 124:9	279:20
95:9,10 109:12	contractual	controversy	125:4 234:19	countermeasu
114:20 119:15	103:12	332:13	304:11	69:3
132:11 133:21	contrary 151:5	convenient	corner 183:22	counterterrism
135:13 144:18	224:3	113:3	183:22 184:1,3	205:3
160:4,20	contrast 31:3	converged	184:4	counterterror
161:15,22	35:22 59:8	335:20	corporate 79:9	6:1,8 109:12
162:14,17	113:12	conversation	79:17 108:13	111:4,14
163:11,21	contribute	135:6 262:13	144:15	114:16 119:9
164:7 168:4	41:21 234:4	277:13 282:13	corporation	144:19 162:14
169:1,10	contributes	362:1 365:5	107:1,4	165:20 166:4
181:21 192:7	362:18	370:3	correct 48:14	204:2,13,14
195:2 204:3	control 5:10	conversations	87:1 181:5	205:7 208:5
205:3 208:3,4	12:13,22 14:1	166:11 274:6	272:9	213:21 216:19
217:3 218:14	15:8 16:4,5	275:10 283:18	corrected	217:6,10
218:19 219:4	17:16,20 28:8	convey 34:19	194:13	218:14 219:18
221:7 276:21	28:8 31:15	36:9	correction 31:6	227:10 246:4
290:16 297:11	37:21 42:20,20	conviction 184:4	55:10 56:7	323:11 325:15
305:19 327:1,3	64:1,7 67:22	convinced 11:12	57:6,13	countertrroris
327:11 328:15	68:2 69:7,20	cook 2:7 4:14	corrections	3:2
332:18 340:20	71:5,7,9,11	88:13,14 92:8	374:13	country 10:22
350:15	78:10 80:17,20	171:10 173:17	correctly 45:20	44:12,22 187:5
contexts 32:18	83:21 87:1,16	175:8,18 196:8	73:11 166:20	187:18 284:11
134:2 160:5	106:5 121:2	203:8,9 206:9	199:11	301:17
304:9 309:13	156:3,4 233:8	207:3 215:9	correctness 87:3	countrys 110:15
continually	293:11 302:15	225:8 237:20	correspond	counts 361:12
24:20 134:3	307:17 308:11	239:10 241:4	34:15	county 385:5
continue 124:8	340:11,12,13	241:14 245:7	cost 50:22 115:9	couple 102:2
	1	l		- F

138:8,9 175:20	373:7,12 374:2	209:18	critize 273:4	342:12 367:20
199:12 224:2	374:13,17	creates 49:11	cross 132:18	cuts 147:15
237:19 249:22	375:6,20 376:1	50:17 175:7	351:1	cyber 111:15,22
273:12 309:18	376:4,10,19,19	305:1 353:12	crossed 172:21	133:3 313:11
319:6 340:22	376:21 377:3	creating 72:17	crucial 114:22	cyberattacks
361:12 381:6	378:5,14,17	135:4 140:17	372:7 375:18	227:13
course 5:15	379:5,14 380:4	158:3 183:20	crumple 40:9	cybersecurities
51:15,19 76:18	380:11,18	244:10 276:18	crumpling	349:22
90:14 99:8	courtappointed	303:19 313:1	307:10	cybersecurity
110:8 119:14	374:22	creation 206:20		295:19 343:11
136:14 150:17		360:17 369:3	cryptographer 134:18	346:5 378:18
173:13 207:22	courtesy 361:20 361:20			
		credit 336:12,13	cryptography	cycle 126:13
209:18 210:4	courts 8:20	creep 366:4 crime 133:3	40:15 91:17 138:2	
214:14 216:14 217:12 218:13	11:13,17 65:18			d 1:14 4:8
	66:18 97:6	183:21 184:1	crystallize 202:5	110:14 124:22
219:9 222:10	130:11 132:12	184:17,17,18	crystallizing	145:19 169:3
222:16 223:16	133:15 364:17	184:20 295:18	148:14	248:1 284:6
239:19 249:4	368:17 372:13	371:15	cues 297:19	daily 34:6
253:17 257:8	372:16 374:6	crimes 183:16	culture 95:1	374:14
258:8,20	375:2,8,17	183:16 184:3	172:1 215:3	dam 351:2
260:11,15	cover 172:13	criminal 53:17	251:19	damage 115:8
264:15 266:4	224:13	65:3 182:12	current 44:19	damaged 128:6
267:8 281:15	covered 14:13	183:5,13 217:3	58:14 101:7	damages 222:22
293:18,19	105:15 349:10	221:9 222:10	189:16 217:13	0
299:5 304:12	383:11	222:22 372:14	231:1 369:14	damaging 161:1 dan 2:15 28:4
309:9 347:6,6	covers 97:2	criminals	currently 70:5	31:22 50:12
347:14 367:6,8	101:20	110:22 111:17	190:1 212:2	65:22 71:2
375:8	cpo 177:8	113:1 182:9	283:8 313:4	
courses 311:17	cracking 187:10	critical 11:1	customary 51:9	98:2,17 103:21
court 12:1,4	craft 87:21	31:16 55:10	customer 125:7	dangerous
14:12,16,18	187:17	89:11 111:14	125:19 127:7	184:5 210:20
54:5 65:5,8,13	cramped 19:11	113:7,21	132:12 150:5	daniel 18:2
71:17 72:1	crazy 354:16	114:15 116:12	172:18 188:6	dans 77:9
74:22 76:8	create 34:18	119:7 125:7	188:14 265:19	dark 185:11
129:8 130:4	47:11 75:13	132:22 187:6	265:21,22	189:21,22
147:16 150:14	76:15 104:19	207:15 224:22	266:9,11,14	darlings 136:13
151:12 196:13	105:3 117:21	232:19 281:5	352:18 353:2	data 18:19,20,21
212:11,21	126:8 132:16	293:11,14,19	customers 125:5	19:1,2,3,4
221:14 245:18	140:5 158:17	300:6 351:9	125:9,15	20:13,17,20
292:9,14	212:9 258:9	352:15 378:13	127:13 131:21	21:14,16 24:3
301:21 314:22	284:4 349:22	379:4	149:8,22 150:7	33:20,22 34:2
325:18,19	353:3 356:22	critically 207:20	150:21 167:21	34:2,3,14,14
358:18,22	created 151:15	criticized 112:9	168:1,2 188:3	35:5,12,15
365:11 369:9	152:8 209:12	183:11,17	188:7 265:4,5	36:2,7,15,22
	1	I	I	I

I 				
36:22 37:14	162:4 164:6,10	305:22 306:3,8	61:2,6 75:4,8,9	deanomoized
38:3 39:7,20	164:10 165:8	307:15,18,19	77:17 79:9	269:16
40:2,2,18,19	168:8,11,15	307:19 308:6,7	102:11 161:3,5	dear 226:21
40:20 43:17	169:7 178:21	308:7,11,12,15	datacentric	death 288:7
46:3 47:3,5,9,9	179:7,11 180:2	308:22 310:4,7	232:9	310:17
49:1,4,10,10	180:8,12,19,21	310:14,14,16	datadriven	debate 96:14
49:13,15 50:15	181:16,21	311:5 312:20	182:22	129:3 180:19
50:16,17 52:2	182:5,16 183:1	317:1,3,5	datafied 312:17	226:11 256:4
52:8,14 55:17	185:18,22	317:1,3,3	datas 52:21	290:14 310:18
II '	· · · · · · · · · · · · · · · · · · ·	· ·	date 37:6 237:16	334:14 337:4
61:22 62:2,3	186:1,5,5	321:22,22		
62:10,13,13,14	187:2 188:4,8	322:5,7,8,8,9	dave 250:19	376:17 378:21
62:21 63:4,7	188:9,12,18	322:10,11,12	david 2:3 10:18	debated 154:5
63:10 68:12	189:3 190:14	323:2,3,3,6,7,9	263:9 285:9	212:2
70:19 76:12,14	190:19,20	323:12,13,16	day 88:17 97:21	debates 59:14
77:2,15,20	200:22 201:3	323:19 324:2,3	101:22 133:1	115:5 131:16
79:13 82:6	228:20 229:21	324:5,10,14,19	150:6 271:7	debating 113:5
87:1,3 90:14	230:19 232:8	325:4,17,22	285:2,12 289:4	decade 242:7
91:6,11 94:4	232:11 233:22	326:3 327:14	290:21 303:21	decades 123:20
102:11 104:17	234:20 235:21	328:1 329:8,11	304:17 315:5	124:21 185:2
105:1,1,2,13	235:22 237:1,2	330:12,16,17	319:16 338:5	340:22 369:13
113:8,10 114:6	237:8 256:8,10	330:18 331:2	383:5,15	december
114:22 117:15	256:13 258:9	333:9 335:1,4	385:13	385:18
117:16,18,19	258:14,16,18	339:4,6 342:12	days 57:4	decentralized
118:7,8,21	259:1 262:17	342:12,22	170:16 248:15	228:9
119:1 120:21	262:17 269:7,8	343:11 348:22	312:17 384:7	decide 51:13
121:15,19,22	269:16 270:3	349:17 350:2	de 53:6	100:14 187:13
122:1,11,16	272:12 275:4,4	350:16 352:22	dead 224:4	269:20 376:22
123:3,11,19	277:20,20	353:19 359:16	deal 93:17	decided 8:21
124:1,6 125:12	282:19 289:13	360:3,6 362:13	100:13 105:15	78:19 273:1
126:1,2,3,4	289:14,15	362:15 366:13	133:22 135:15	decision 74:17
127:8,14,15,19	290:16 291:10	366:16 367:3	155:15,15	133:15 145:22
128:7,10,12	292:2,7,9,22	367:18,19	169:1 191:1,4	221:13 245:3
129:17 130:6	293:4,9,10	368:10,10,10	191:9,18 239:7	257:13,14
130:18,21	294:13 296:16	368:14 370:13	270:8 306:2	261:17 379:21
131:2 132:5,13	296:20 297:6,7	371:5,7,8,10	325:16 343:14	382:4
132:18 139:10	297:9,15,15,20	371:13,18	dealing 88:19	decisionmakers
139:16,18	298:3,11 299:7	372:3 381:21	137:9 150:5,6	253:12
140:6,6 141:10	299:8,11,21	382:12,14	155:19 171:20	decisionmaking
147:20 149:17	300:3,15,17,19	383:15	247:17 305:20	116:8 349:6
150:8 151:14	300:20,22	database 60:14	310:15 349:16	decisions 5:1
151:14,15,20	301:7,7,18	79:17 160:17	349:20 374:21	48:18 49:4,5,6
152:1,8,17	302:5,10 303:6	195:2,5 248:20	dealings 103:1	57:20 115:5
153:5 154:3	304:4,5,5	248:22 365:11	dealt 72:18	140:13 145:7
156:2 160:15	305:19,20,21	databases 59:17	177:7	151:12 242:13
100.10			- · · · · ·	

252:5 309:22	339:19 368:12	27:16 33:4	3:13 25:8	342:15 351:21
340:19 348:21	definitions 5:13	66:6 253:13	100:5 122:5,6	deserve 42:6
350:2	11:9,11	295:16 296:7	122:15 123:16	deserves 14:22
declaration 51:6	definitive 100:3	democratic	183:10 209:15	362:10 365:2
declaring 51:14	287:11	54:13 55:3	213:1 215:11	deserving 43:6
declassification	degree 59:7	56:1,6 302:22	215:19 216:2	design 61:16
301:21	103:13 119:22	demonstrate	217:4,18 222:7	125:17 126:17
declassified	254:1 345:5	266:16	225:5 226:2	135:17 142:12
129:6 377:15				
	347:10	demonstrating	246:3,7 247:1	142:13 143:11
decrypt 112:6	degrees 8:16	284:11,12	290:4 303:13	143:17 144:8
dedicate 319:9	196:11	dempsey 2:6	313:11 364:15	144:15 146:15
dedicated	deidentification	4:14 96:7,8	departments	154:18,20,22
119:13 126:21	111:10 113:5	97:11 98:19	215:15,18	155:16,17,17
193:22	114:4,11,13	99:16,20 101:9	216:20 217:1	158:19 190:2
deemed 57:7	115:17 160:3,8	102:1 109:6,8	219:19	designed 209:21
deeper 172:14	175:16 194:7	117:2 124:10	departmentwi	211:19 220:17
198:22 199:4	195:9,18 196:9	134:8,11	216:3 239:19	232:13 233:9
deeply 40:22	383:13	141:16 144:6	depend 269:6	365:22
95:22 247:2	deidentified	144:12 145:12	371:22 372:1	designs 134:16
default 112:4,13	194:10,12	145:15 146:12	dependent	desirability
115:16 166:14	196:16	146:17,21	237:9 297:1	214:8 294:7
185:20 187:4	deidentify	149:5 150:11	332:16 377:9	desirable 374:15
359:8	195:14	151:21 153:7	depending	desire 4:22
defend 244:5	delegated 85:15	154:11 158:8	288:12 360:5	254:20
defense 122:6	253:10	158:22 159:14	depends 58:10	desired 38:10
123:16	deleted 307:8	164:20 165:14	102:19 162:9	desk 306:14,15
define 7:22	358:6	181:8 183:7	359:18,19	despite 14:14
138:4 140:5	deletion 307:9	184:22 185:4	deploy 78:21	38:6 234:21
155:5 171:6	307:12	187:21 189:7	deployed 127:4	destroy 167:2
208:2 269:10	deliberately	192:1 201:8,11	183:10 304:14	destroyed 222:1
269:14	68:9	202:21 278:4	309:12	300:19
defined 45:17	delineated 53:16	347:17,19	deployments	destruction
45:18 219:2	65:12	350:6 351:5	304:8	118:8 365:17
223:12	delinked 196:12	354:2	depot 141:11	detached 374:17
defines 42:1	delivering	denial 112:14	depths 237:16	380:4
defining 1:5	265:18	denied 60:4	deputy 3:20	detail 175:12
2:10 4:4 5:21	demand 24:8	221:12	25:7 215:16	218:18 220:22
51:15 202:3	188:7,14	denominator	313:6	237:15
203:6 204:3	demanding 24:9	46:9	derive 320:12	detailed 34:22
definitely	demands 129:13	deny 59:17	derived 227:6	75:14 122:15
104:11	189:6	77:10,10	derives 26:22	129:12 130:1
definition 11:15	democracies	294:18	345:14	308:17 309:2
11:19 157:16	32:10	denying 336:13	describe 227:2	details 151:16
208:18 301:4	democracy 3:19	department	described	detain 298:11
200.10 301.1				270.11

				1
detect 339:3	155:20 156:4,7	39:12,12 47:4	235:11 238:20	250:18 252:10
detection 208:16	156:20 185:21	48:3 63:18	269:3 287:17	295:15 313:6
deter 192:15	290:8	68:10 79:3	290:10 330:4	directories
determination	devices 39:16	89:17 94:10,11	368:17	194:15
258:22 300:20	136:2 137:9	96:17 97:2,3	difficulty	disagree 89:16
358:7 359:6,8	138:5,14,16,18	103:11 157:16	374:19	296:13
359:12 377:20	138:19 139:4,8	161:4,5 165:1	diffusion 186:11	disagreement
determinations	142:5 143:12	172:17 173:10	186:12	25:18 26:13
258:3	155:11 158:4	177:9 184:9	diginity 231:8	disappear
determine	185:16 188:5	202:13 212:15	digital 49:2	156:17
114:10 235:6	290:6 329:16	216:15 230:13	136:10 234:10	disappearing
235:15 236:3	devolve 277:14	232:10 235:16	diligent 14:15	140:8
276:3	dhs 205:17	235:19 236:4	207:18	discard 101:6
determined	250:13 271:16	236:12 238:6	diligently 246:8	discern 190:22
34:14 113:18	272:3 273:5,17	241:5 244:19	diminish 52:18	discharge 123:4
determining	276:2 282:9	244:20 248:13	147:9	disciplines
40:19	326:17	249:3,5 250:17	diminishing	237:10
deterrent 325:8	dhss 273:4	251:1,7 252:2	192:13 319:12	disclose 34:6
325:13	dialogue 154:6	259:19 260:14	diminishment	99:3 102:20
detrimental	313:14 365:8	262:11,13,15	52:19	129:12 266:5
161:6 166:3	dias 216:21	263:7 265:14	dinner 161:14	280:18
develop 39:3	didnt 13:14	273:8,12	diog 217:21	disclosed 79:11
114:18 126:18	81:18 136:21	277:20,20	220:4,8 221:17	209:22 222:19
148:6 173:2	147:1 167:10	297:22 298:9	221:18	discloses 150:16
235:6,10 236:6	185:10 201:6	304:20 309:13	direct 102:5,6	disclosing 34:8
236:8	243:22 271:2	314:7 315:22	115:9 280:21	91:18 142:6
developable	326:14 347:13	316:20 319:4	299:7 318:9,19	disclosure 5:7
92:6	352:11	319:19 323:15	326:10 378:10	13:19 81:19
developed	differ 93:20	323:15 327:21	directed 102:10	84:9 97:16,17
114:20 117:14	248:18,19	334:6 339:21	228:4 271:12	97:19,20 98:3
127:4 148:8	376:11	359:9,13	381:10	98:8,8,21 99:5
173:1 177:12	difference 43:19	373:12 381:2	direction 203:16	99:6 222:16,21
234:13	44:3,3,5,5	differentiating	328:21 331:22	disclosures 16:9
developing	52:19 73:16	299:2	369:11	16:11,14 57:10
54:17 91:22	199:12 327:22	differently 21:7	directives 243:1	120:6 163:14
92:10 153:14	differences	21:8,8 259:18	directly 30:17	222:11 278:21
154:8 234:8	264:13,16	differs 93:19	34:6 43:10	discontinue
263:19 276:9	299:14,19	difficult 36:8	206:12 227:14	32:6
development	different 6:10	37:20 38:12	276:1 300:15	discount 69:14
5:3 126:12	10:16,17 11:9	86:18 138:5	307:3	discourse 17:3
developments	11:21,22 12:3	143:2 162:2	director 3:11,20	discoverable
129:5	18:18 21:5	163:7 169:16	112:9 117:4	229:2
develops 225:19	27:8 28:16	169:22 170:1	206:6,12	discplinary
device 136:15	32:18 33:15	214:10 233:21	225:15 226:1	281:21
		•	•	•

				7
discreditable	displayed 43:20	documented	266:5,17 267:7	120:15 123:1,2
18:16	disposition	331:16 354:1	268:4 271:1	132:21 137:17
discreet 89:6	222:3	382:4	274:11 277:7	143:3 144:9
discrepancies	dispute 58:6,7	documenting	277:10 279:22	148:22 149:20
269:9	disruption	233:7	282:10 283:8	152:4 155:4,21
discrete 335:1,4	46:14	documents	283:16,17,22	157:13,14
346:1	disruptions	218:1,10 220:3	284:12 294:7	158:10 160:20
discrimination	45:21	doesnt 13:9	297:6 319:13	162:13 163:5
123:7,21	disrupts 47:18	15:21,22 20:1	321:18 322:4	165:4 166:13
discuss 9:5 10:1	dissemination	20:6 23:9 36:9	326:2 330:3	166:13 170:1,2
36:6 63:11	323:3 324:12	42:21 46:21	331:15 332:3	171:1 173:11
105:11 205:2	325:6 351:10	50:2 52:7 59:4	334:1 342:1	178:5,22 179:2
237:15 280:15	dissimanation	65:11 69:12	343:3 351:15	182:3,21
discussed 79:1	258:5	71:22 72:11	351:20 354:17	188:18 189:11
119:2 154:4	dissonance	80:12 82:9	358:8 366:21	189:12,22
204:7 210:1	108:8	93:17 97:22	379:9 382:18	190:9 191:4
212:2 226:16	distinction	100:15 149:20	doj 219:16	193:9,14,16,17
234:1 358:18	266:3 336:6	151:6 163:15	224:9,18 239:1	194:2,3 196:15
discussing 220:8	distinguish	167:14 192:15	239:20	198:6 201:7,13
290:16	380:9	197:14 218:22	dollars 141:8,9	218:4 220:21
discussion 5:15	distrust 349:8	248:3 258:17	domestic 113:22	223:22 239:2
7:3 10:13	distrusting	266:11 271:21	160:4 161:22	241:18 248:7
25:14 27:13	155:21	272:14 282:1	217:19,21	249:16 253:13
29:7 63:14	disturbed 5:5	294:6 315:4	365:16	254:10,14
119:11 124:17	diverse 33:4	323:12 328:7	dominant	256:11,16
124:18 145:6	dividend 190:7	356:1 362:20	144:16	258:19 261:15
155:4 197:13	division 216:13	364:6 381:1	dont 11:17	267:10 270:21
213:4,6,11	219:19 223:3	doing 22:12	13:22 15:14	272:20 273:4
215:7 234:4	246:19 250:16	23:7 52:9 68:8	17:8,11 21:8	274:4,7,9,21
235:13 268:12	doable 140:17	91:7 95:17	22:10,13,17	274:22 275:14
297:5 353:20	doc 99:13	116:3 136:1	23:7 25:22	275:16,18
383:8	doctor 98:22	139:15 174:10	26:3,16 43:14	276:12 277:2
discussions	99:5,9,11,14	177:15,20	44:9 47:20	279:6,8 283:19
131:22 153:20	332:20	191:12 198:14	48:6 49:22	284:7 288:16
165:11 235:19	doctors 99:4	211:7 238:6,21	51:12 53:5,10	289:21 290:20
338:20	doctrine 13:21	241:8,17 242:6	55:7 56:16,22	291:6 297:3
dishonorably	72:5 97:13,13	242:22 243:4	58:5 59:16	307:21 309:19
123:4	97:22 151:6	243:12,21	64:8 68:2	309:20 311:15
disincentive	168:17 197:13	244:20 245:14	70:13,19 73:12	317:22 318:16
292:12	353:15	249:8 250:13	78:2,13 79:14	328:10 329:17
dismissed	document 272:4	251:12,20	82:12 83:6	329:18 332:8
353:14	324:17	257:3 261:1,2	101:1,19	332:18,20
disparate 96:17	documentation	261:3 262:4	102:18 104:13	334:4 335:8
121:11	249:10	263:4 265:18	106:15 107:16	336:20 338:11

339:18 342:11	driving 56:10,14	ecommerce	310:13 356:8	345:20
344:2,16	117:22 188:14	206:17	effectiveness	elephant 201:14
345:11 349:9	drones 15:6	economic	319:20 377:2	eliminate 56:16
354:3 356:20	369:16	138:17	effects 36:12	56:22
357:1 359:12	dropped 16:20	economy 289:19	37:2 291:8	elisebeth 4:14
360:9,14 361:2	duck 279:6	291:20	312:22 366:14	elizabeth 2:7
367:14 371:21	duckduckgo	ed 2:12 33:11	effectuate 11:2	elses 51:18
376:8 377:5,6	198:13	75:12 77:11	efficacy 213:20	elusive 235:2
377:19 381:10	dupree 7:15	87:10 101:11	380:6	email 132:12
381:13,14	286:15	105:10 115:21	efficient 294:21	188:21 343:17
door 14:3	dust 75:1	182:4	356:19	343:19,22
113:16 120:13	dusty 306:16	edge 270:9	effort 87:2	embed 354:13
173:3 194:20	duties 205:14	eds 86:10	118:12 323:12	356:13
doortodoor	dynamic 308:2	education 225:9	efforts 7:17 32:6	embedded 30:2
14:16	dynamically	311:10 356:15	207:19 217:2,8	127:1 218:6
dossier 49:2	348:20	357:2	217:9 238:14	220:5 224:5
doubting 351:14		educations	egaurdian	230:8
369:11	E	316:12	220:21	embodied 80:21
doubts 243:3	e 229:22 230:1	effect 26:6 34:19	egovernment	embrace 152:15
downsides	e0 278:22	35:14 36:14,18	240:21	emergency
282:22	ear 164:20	66:10 79:2	eguardian	99:12
downstream	earlier 57:15	80:12 168:5	220:15 221:2	emerging
36:12 334:7	79:2 101:21	169:2 315:17	eight 288:11	171:20 197:6
335:6	129:15 142:22	325:9 330:3,19	either 21:5 61:4	290:14
downtown	144:21 161:2	360:22 379:12	62:6 66:11	emitting 142:5
248:1	179:22 194:8	effective 35:17	67:1 72:21	emphasis
dozen 360:19	194:19 199:19	41:8 67:14	77:13 98:13	100:16 101:12
dr 160:1 189:10	200:4 203:10	69:19 83:9	103:5 115:9	300:3
drafters 53:13	226:16 246:1	93:1 94:20	183:15 245:3	emphasize
64:21	310:5 365:8	114:4,12	261:14 264:7	113:8 327:15
dragnet 67:14	early 143:18	130:12 171:18	264:22 281:6	emphasizes
draw 184:2	easier 38:18	173:21,21	282:3 283:20	108:1,14
280:10	39:1 112:1	174:14 188:21	291:22 294:6	employ 326:22
drawer 306:15	141:4 145:10	192:18 208:12	328:8 351:14	employee 95:12
drawing 255:20	324:22 325:7	210:11 213:22	364:16	343:14,17,18
drawn 259:18	349:15	246:15 276:19	elaborate 38:21	343:21
draws 102:11	easily 75:1	283:5 292:17	69:2 121:6	employees 95:17
278:16	162:4	311:7 312:18	162:18 217:4	107:14 126:8
dream 303:21	easy 133:12	312:21,22	electronic 290:6	172:13
drive 57:1 63:17	193:18 328:22	319:18 321:12	308:10 319:2	employer 339:6
148:19 172:10	349:18	321:21 322:15	element 64:1	339:16
382:3	eat 86:15	342:7 366:19	128:22 323:19	employment
driven 188:6	eavesdropper	effectively	324:10 346:11	37:13 59:17
229:16 336:16	39:10,13,17	118:12 310:1	elements 231:17	60:4 293:1
	<u> </u>	<u> </u>	I	I

			Δ(
292.12	100.15	73.14.220.22	310:6
			erosion 115:12
\mathcal{C}			errand 381:5
			error 31:6 55:9
			56:7,17,19
			57:1,6,13
			58:16 59:21
		,	84:16 87:5
		,	174:15
			errors 56:11,14
			87:1
			escalation 127:3
•			especially 37:16 61:7 63:22
			121:22 133:10
3			143:9 155:18
			180:9 192:6,10
			280:5 290:15
· · · -			310:15 322:14
	\sim		328:17 338:16
			349:16 357:9
		'	essence 233:3
	• •		essente 255.5 essentailly
			265:21
			essential 25:16
		-	31:14 59:10
		_	64:2 111:8
· ·		· -	125:5 316:12
		_	346:13 374:7
		_ ~	essentially 25:17
	· · · · · · · · · · · · · · · · · · ·	_	74:17 90:3
		-	168:17,19
			169:21 195:9
			314:11,16
cinoi cing	220.20 232.13		· ·
140.15 342.6	241.6 6 246.9	228.15	315.19 22
140:15 342:6 engage 248:8	241:6,6 246:9 ensuring 209:5	228:15	315:19,22 316:7 9 324:3
engage 248:8	ensuring 209:5	equipped	316:7,9 324:3
engage 248:8 276:14 319:5	ensuring 209:5 245:20 295:11	equipped 377:16	316:7,9 324:3 324:15 325:2
engage 248:8 276:14 319:5 363:15,18	ensuring 209:5 245:20 295:11 296:2	equipped 377:16 era 186:4,4	316:7,9 324:3 324:15 325:2 326:2 335:16
engage 248:8 276:14 319:5 363:15,18 engages 218:12	ensuring 209:5 245:20 295:11 296:2 enterprise 274:4	equipped 377:16 era 186:4,4 228:2 234:20	316:7,9 324:3 324:15 325:2 326:2 335:16 335:17 345:16
engage 248:8 276:14 319:5 363:15,18 engages 218:12 281:17	ensuring 209:5 245:20 295:11 296:2 enterprise 274:4 enterprises	equipped 377:16 era 186:4,4 228:2 234:20 erika 3:13	316:7,9 324:3 324:15 325:2 326:2 335:16 335:17 345:16 345:21 360:22
engage 248:8 276:14 319:5 363:15,18 engages 218:12 281:17 engaging 52:8	ensuring 209:5 245:20 295:11 296:2 enterprise 274:4 enterprises 125:10	equipped 377:16 era 186:4,4 228:2 234:20 erika 3:13 215:10 225:8	316:7,9 324:3 324:15 325:2 326:2 335:16 335:17 345:16 345:21 360:22 362:6,22
engage 248:8 276:14 319:5 363:15,18 engages 218:12 281:17 engaging 52:8 72:15 199:3	ensuring 209:5 245:20 295:11 296:2 enterprise 274:4 enterprises 125:10 enterprize	equipped 377:16 era 186:4,4 228:2 234:20 erika 3:13 215:10 225:8 239:10 244:17	316:7,9 324:3 324:15 325:2 326:2 335:16 335:17 345:16 345:21 360:22 362:6,22 365:11
engage 248:8 276:14 319:5 363:15,18 engages 218:12 281:17 engaging 52:8	ensuring 209:5 245:20 295:11 296:2 enterprise 274:4 enterprises 125:10	equipped 377:16 era 186:4,4 228:2 234:20 erika 3:13 215:10 225:8	316:7,9 324:3 324:15 325:2 326:2 335:16 335:17 345:16 345:21 360:22 362:6,22
	383:13 endangering 200:11 endeavor 56:18 ends 108:16 361:18 endurance 285:1 enduring 133:16 enemies 44:14 55:18 115:10 enemy 114:7 160:2 energy 347:7 enforce 135:12 155:5 158:4 174:13 enforceable 135:5 140:14 140:17 158:6 169:11 enforced 171:22 enforcement 38:22 69:10 70:16 112:7,18 132:10,22 133:20 168:4 169:1 189:16 191:6 217:3 219:18 220:18 221:7 224:14 282:3 304:6 enforcing	endangering 200:11engineer 176:21endeavor 56:18303:10 304:18ends 108:16 361:18357:11,11,12endurance 285:126:4 305:6enduring 133:16143:15 172:15enemies 44:14 55:18 115:10305:11 309:20enemy 114:7 160:2311:14,15,18energy 347:7 enforce 135:1231:22 338:18155:5 158:4 174:1331:22 338:18enforceable 135:5 140:14 140:17 158:6 169:11215:2 256:2enhance 210:16enhance 210:16enforceable 135:5 140:14 140:17 158:6 169:1127:13 144:7enforced 171:22 enforcement 38:22 69:10 70:16 112:7,18 132:10,22 133:20 168:4 169:1 189:16 191:6 217:3 219:18 220:18 171:19 173:4126:4,5 133:4171:19 173:4 221:7 224:14 282:3 304:6 enforcing173:11 192:15	endangering engineer 303:10 304:18 301:17 endeavor 56:18 357:11,11,12 entirely 25:13 ends 108:16 26:4 305:6 39:8 42:19 endurance 311:17 entities 24:4 285:1 engineers 127:5 enduring 143:15 172:15 209:14,18 133:16 249:4 304:22 entity 102:15 enemies 44:14 305:11 309:20 entity 102:15 55:18 115:10 311:14,15,18 106:8 230:1 entity 102:15 55:18 115:10 311:14,15,18 302:15 340:14 entitys 102:15 60:2 356:16 374:19 enhance 210:16 entitys 102:14 entitys

established	76:3 77:4 85:4	160:8 161:21	exemption	152:21 157:16
210:8 313:13	98:18 137:6	170:9 173:18	14:13	166:19 191:22
establishing	241:22 270:5	174:6 175:15	exercise 15:9	197:7 198:7
135:9	274:1 289:20	180:18 195:12	192:19 207:19	334:11 362:4
establishment	304:3 320:17	196:3 210:7	210:5 276:15	363:7,10,20
89:10	321:8 353:19	212:6 218:9,20	328:12 340:14	364:3,18,21,22
esteemed 140:12	367:2,18	229:10 231:19	exercises 192:12	366:5,17
estimate 164:4	everyday 36:1	239:11 240:19	325:4	368:13,20
etcetera 19:8		243:14 245:18		369:2,6,12,15
54:5 64:9	everyones 147:18 195:4	250:6 269:15	exercising 107:11 209:16	369:2,6,12,13
91:10 153:4	285:1	270:5 271:17	exfiltration	expectations
169:7 178:11	evidence 67:12	281:17 290:4	343:12	47:18 147:9,11
188:5 254:12	221:19,20	301:3 320:7	exhaustive	149:13 150:9
279:10 313:1	222:3	329:3 343:9	380:16	151:19 167:3
348:6	evidentary	349:18 351:17	exist 55:16	196:20 226:12
ethical 172:13	272:15	353:12 369:5	80:12 136:21	323:15 344:15
379:18	evidentiary	examples 35:19	167:10 309:8	346:18 347:4
ethicists 237:12	221:22 275:2	35:20 36:3	existed 136:12	380:14
ethics 294:6	evolution	58:13 122:4	309:6	expected 47:14
311:17	124:14	exante 70:18	existing 71:16	148:7 212:14
ethnic 122:1	evolves 134:3	excellent 66:17	128:18 132:19	expecting 98:18
ethos 335:12,14	149:16 363:8	171:11 296:1	233:14 281:8	experience
europe 290:15	evolving 196:19	exception 14:8	exists 158:14	98:17 154:18
343:5	196:20 226:10	99:10 168:18	302:22 304:18	216:11,11
evaluate 23:6,12	exact 275:6	exceptions	expand 115:19	242:17 243:5
24:21,22 225:2	exactly 57:4	53:16 65:4,12	149:4 167:11	244:16 314:15
236:10	66:10 77:9	65:15 71:21	198:9 282:8	317:16 321:4
evaluates 73:4	88:19 89:13	295:3 344:1	368:12	325:12 347:2
216:1	147:14 214:14	excessive 365:19	expanded 188:2	experiment
evaluating	307:7 308:11	exciting 226:8	217:15 304:7	54:19 55:14
235:7 296:6,19	308:13 310:11	exclusion 48:11	306:7,10	experiments
332:1	310:13 320:20	exclusive 309:11	expands 231:15	250:1
evaluation	339:18	exclusively	expansions	expert 61:11
263:16 332:10	examined	131:22	217:16	225:14 256:7
333:8	281:12	excuse 145:14	expect 5:14 36:2	380:11
event 7:17	example 27:12	317:22	40:7,8,8,9	expertise 126:22
137:17 208:8	29:4 34:15	executive 117:4	119:22 148:19	184:8 244:15
events 123:6	35:7,19 37:4	127:9 193:1	148:22 149:10	249:20
323:1	38:18 39:8	212:22 227:9	149:12 166:22	experts 41:10
eventually	52:11 59:1	229:18 279:1	167:9 191:13	110:16 178:12
379:14	82:5 86:22	301:5	366:15 368:14	237:12 249:14
everybody	99:8 113:22	exemplifies	expectation	374:22 376:15
51:18 53:11	126:11 145:12	301:12	74:17 147:6	378:7
64:7,16 67:18	145:15 148:2	exempt 222:13	148:14 149:4	expired 80:13

205.10	245.17.260.11	110.16 110.1 7	246.5 205.17	f ood il 1 0 4 2
expires 385:18	245:17 260:11 266:10 276:4	118:16 119:1,7 120:20 122:4	246:5 305:17 faith 256:20	feasible 84:3
explain 22:10,13 147:13 160:7				92:11 115:13
	302:4 340:18	133:2 143:10	fall 65:12 95:6	181:19,22,22
218:5 283:22	363:8 378:5	143:19 145:1	300:10 344:15	359:17
284:4 375:10	external 244:14	147:10,10	falls 43:5 180:20	feature 365:10
378:7,15	346:12	153:8 170:6	false 5:9 56:12	374:8
explained 248:6	externally	175:11,12	56:12,13 58:16	features 175:11
375:4	346:16	177:19,22	familiar 30:8	188:15
explicit 98:8	extra 87:4	184:10 190:13	304:2	federal 4:10
345:16	177:16 205:21	193:11 197:3	familiarity	30:8 122:11
explicitly 34:8	294:19	197:14 198:2	377:21	132:11 205:3
252:18	extract 91:11	199:6 212:18	families 122:9	216:12,14
exploited 378:1	extraordinarily	213:7 223:10	famous 35:7	220:19
exploiting 39:17	117:15	223:12 224:12	fan 190:11	feedback 183:20
exploits 111:20	extreme 108:12	243:10 258:2	far 9:6 28:13	211:12 352:18
166:9 170:7	115:8	279:14,19	29:11 75:17	feel 29:9 41:20
exploration 21:4	extremely	280:4,17 292:8	123:21 168:10	54:18 128:10
explore 7:21	108:20 161:1	292:18 315:9	218:4 278:14	177:22 207:11
339:8	197:12 204:5	318:2 332:7	288:14 291:2	240:8 243:1
exploring	207:14 276:21	346:19 352:2	317:19,20	260:20 293:10
285:15	361:6 378:2	373:18 376:13	fast 277:22	355:8,21
explosion 138:3	eye 105:5	factored 177:11	371:22	feeling 16:6
expose 49:12	eyes 364:19	factors 94:19	faster 117:18	54:20
171:2 228:22	370:22	250:9	138:10,11	feelings 51:11
exposed 47:20		facts 5:7 34:3	169:13	fellow 9:18
47:21	F	35:6 47:8 48:3	fathers 8:11	25:19 74:11
expost 70:17	face 26:1 36:9	factual 219:1,9	fathom 139:19	206:2 252:16
87:4	100:19 268:15	fading 150:10	favor 4:17	286:10
express 30:16	facebook 17:10	fail 37:22 38:9	384:11	fellowships
327:22	78:15 136:21	fails 299:16	favorites 136:6	311:11
expression 21:3	191:17 197:2	failure 38:5	fbi 81:4 112:9	felt 243:11,17
expressly 204:3	198:21 199:1	58:12	217:8,19 218:1	269:11 339:12
exquisite 380:8	339:1 352:19	failures 38:3	218:12,19	377:11
extended 15:2	facial 148:3	58:6 95:15,19	223:3,15	felten 2:12
304:3	facilitates 129:2	fair 67:16 100:2	225:10 239:11	33:11,17 60:22
extensive 7:2	facing 344:6	100:17 125:20	239:16 240:2	61:19 63:2
74:1 230:4	fact 13:22 14:14	134:6 149:7	240:13 241:3	67:21 79:6
extent 12:14	19:22 24:15	204:8 225:14	246:19 248:19	80:2,5,7 86:16
52:16,17 86:16	25:15 37:7	231:20 296:8	281:14 306:13	89:20 90:13
89:5 123:10	43:14 48:11,22	301:19 303:3	fbis 217:14	95:15 105:10
142:17 151:5	49:6,13 71:20	336:12	225:9 252:3	105:19 115:21
174:12 188:13	75:6 78:8,14	fairly 58:15	fear 28:13 56:2	169:17 200:4
196:19 213:21	79:16 88:18	152:10 240:21	382:14	fewer 320:22
213:22 222:14	98:19 99:2	244:21 245:9	feared 16:15	fiction 13:20
		= : ::= = = :::		1

80:9,11,11	find 26:12 38:7	350:8 354:8	288:21 297:13	flick 348:16
field 61:13 64:3	117:19 161:6	370:3,7 381:11	300:2 313:12	flip 371:17
155:18,20	184:12 215:1	381:13,14	313:17 314:11	flipped 359:15
176:10 223:7	241:8 244:20	382:19 383:12	318:20 322:17	flow 105:13
249:14	257:5,6 267:21	fippslike 289:15	323:7 326:10	142:15 143:21
fifteen 149:6	271:1 311:6	316:3	333:12 344:22	143:22 156:4
fifth 131:14	318:11 319:16	fips 220:1 221:5	345:13 346:3	flowing 185:22
294:8	345:11 347:9	222:18 224:3	348:7 352:5	flows 40:16 68:6
fight 133:3	349:4 361:8	231:21 232:1,7	358:15,16	186:6 242:2
217:2	finding 116:11	233:13 260:17	361:13 364:5	308:11,11
fighting 111:9	fine 88:5 105:18	261:11 271:9	381:12	319:2
figure 62:10	159:6,7 182:18	271:15,16	fisa 129:8,17	fluid 237:8
84:22 123:20	367:2	272:3 273:1,15	130:4,11,14	flush 359:7
159:10 162:11	fingerprint	274:21 277:3	179:15,20	flushing 360:8
190:7,8 214:3	170:11,13,20	282:1	229:19 245:18	flying 194:2,3
261:22 262:1	170:11,13,20	fire 357:14	301:21 352:12	focal 238:11
313:21 316:9	fingers 172:21	firewall 179:5	358:18,22	277:12
316:16 340:4	finish 324:20	firm 124:22	373:7 376:21	focus 5:21 12:10
379:19	fipp 326:15,22	248:1 267:15	fisc 164:18	38:13 46:13
figured 191:11	327:3,10 328:7	first 5:21 7:18	165:3 353:13	94:19,20 95:14
254:16	328:14	13:3 18:7 34:1	374:10	111:6 120:3
figuring 191:13	fipps 25:21 26:5	34:5 36:6	fisma 320:17	131:22 150:3
191:15 312:21	80:21 81:7,8	55:10 61:20	fit 58:9 62:7	217:7,8 228:3
file 34:22 74:22	82:2,7,12 83:5	62:7,14 65:17	97:22 206:7	230:18 238:14
151:10 152:14	83:17,18,22	71:4 76:18	322:11	243:12 244:12
222:2 306:13	84:1,6,12 86:3	80:16 94:22	fits 50:1 101:19	244:14 256:2
306:19,22	86:17 87:20	99:22 102:9	261:10,11	258:21 264:9
307:5,7,8	88:3,20 99:22	110:11 111:7	five 4:12 18:6	268:18 289:16
filed 129:10	100:8 101:18	114:18 118:20	76:1 139:7	289:21 290:17
files 34:14,17,19	118:6,6 119:4	119:2 120:4,8	141:20 159:16	292:1 293:12
35:3	119:7,14	121:5 124:22	177:1 235:10	294:12 305:13
filings 375:12,12	152:17 162:17	128:7 142:1,9	278:9 286:10	323:18 361:15
fill 334:12	271:18 288:3,8	142:12 148:13	322:9	361:22 362:8
377:18	288:12,19	158:8 159:22	fix 57:12 305:2	382:2,6
filled 283:13	289:7,21 290:1	162:20 171:12	305:7	focused 31:5
filter 188:22	291:18 296:18	179:20 186:13	flannel 164:22	127:18,20
final 6:6 39:21	296:22 297:8	189:9 199:10	flashlight	128:11 131:15
134:14 285:3	297:16 299:17	206:21 226:6	350:15	131:17 177:3
339:19 372:16	299:20 300:4	228:1 231:19	flat 136:21	209:5 241:21
finally 7:13 95:5	326:8,14,16,19	235:14 242:1,7	flatirons 311:12	242:20 244:6
114:15 119:21	328:19 329:4,4	242:19 253:22	flaw 253:1	248:2 260:19
145:22	330:2 331:8,20	255:21 274:4	flawed 71:18	285:19 290:6
financial 304:9	332:3,8,15	276:1 281:19	flaws 83:5	292:14 293:7
336:21	333:4 335:13	285:13 287:13	flexibility 38:15	296:16 352:9
		I	I	I

25251120				
353:7,11,20	372:12	forward 41:14	293:13,14,18	22:20 27:13
focuses 12:15	forces 346:22	96:15 116:7,22	294:3,5 314:12	212:6 266:21
295:17 338:3	376:18	141:1,15 215:6	350:12 353:12	freedoms 133:9
focusing 211:6	forcible 120:20	269:20 270:10	353:14 368:22	freely 22:18
224:14 281:14	forcing 263:15	305:16 352:16	372:15 383:12	127:13
341:20 350:9	268:4	383:18	framed 229:20	frequently
350:10 382:20	forecasting	foster 29:7	366:1	228:13 238:2
foia 14:6,8	381:4	127:16 129:20	framers 313:19	238:22 248:16
folks 178:20,22	foreign 111:16	130:4	framework 46:1	288:14
193:19,21,22	115:3 179:15	fostering 27:9	82:13,17 100:8	friedman
196:22 203:10	212:10,21	fosters 27:1,17	100:9,12,20,21	136:19
205:19,21	227:6,7,16	found 88:15	132:17 154:9	friend 13:8,14
235:20 238:22	228:3,5,12,16	92:22 171:18	205:22 231:21	friends 17:11,12
251:14 255:3	229:18 232:6	173:20 235:1	232:1 274:3	70:1
269:3 283:15	256:12 301:5	246:22 256:9	296:10,19	frog 347:10,12
285:20	358:7 381:21	316:13 323:8	297:1 326:20	front 6:22 9:11
follow 65:8	foremost 313:17	foundation	327:12,17	14:2 85:21
126:8 173:17	322:17 344:22	135:9 137:20	332:1,15 333:5	109:22 145:6
189:13,13	346:9 361:13	foundational	333:6 338:12	203:17 207:6
205:21 211:18	364:5	54:9	368:21 369:7	241:10 254:10
211:22 212:14	forget 23:16	founded 23:4	369:12,21,22	286:3 303:21
213:8,9 232:8	form 195:17	25:20	370:4 372:5,7	365:20
298:6 299:17	225:4 233:1,19	founder 25:5	379:21	frontend 245:16
followed 9:17	236:16 242:9	33:13	frameworks	300:6,14 359:2
12:7	270:14	founders 304:16	204:8 231:16	359:3,16 371:9
following	format 9:9	founding 8:11	296:12 333:7	372:6
138:20 140:11	former 3:20	four 5:20 118:19	framing 265:14	fruit 305:17
185:15 211:20	211:21 313:6	236:21 313:15	framwork	frustrated
212:18 213:5	373:4	317:14 357:15	211:18 237:1	373:13
242:22 260:15	formerly 25:7	fourth 12:20	frankly 56:11	frustrates 38:4
344:21 363:3	122:6 376:19	14:22 22:21	288:3 289:18	38:5
follows 36:19	forming 145:2	27:22 29:13	290:1 319:3	frustrating 86:7
43:10 98:12	forseeable	42:2,14 43:6	325:8 329:10	ftc 264:8
followup 245:7	221:22	52:9 53:7,14	franzen 8:3	fulfill 300:18
263:14 354:6	forth 268:17	64:13,22 71:15	fred 3:18 350:6	fulfilled 254:7
377:6	284:15 318:16	71:17,22 72:5	350:6	full 22:11 76:3
food 276:22	374:11	72:6,8,20	free 20:9,9 22:6	116:20 123:10
fools 381:5	forthcoming	80:19 81:8	22:6 41:21	156:3 168:21
footprint 246:6	108:21	82:1,2 103:18	54:19 172:3	205:6 209:6
footprints 240:2	fortunately	107:4,9 119:21	241:14 355:7	351:19
force 51:9 61:21	288:8	121:5 130:16	364:7	fulltime 244:12
89:21 90:3	forum 1:5,12	167:13,19	freedom 5:4,4,5	248:8 249:20
195:5 263:21	205:1 224:12	168:18 229:16	5:8,9,12 11:2	fully 38:1 84:12
forcefully	264:3	292:5,13	11:16 14:6	100:11 189:20
	1 20	2,2.5,15	11.10 1 1.0	100.11 107.20

				1
191:14 208:17	gathered 22:4	158:15 191:11	64:14 270:5	373:20 374:3
210:16	48:4 49:15	191:21	giving 87:11	380:20,21
fun 275:9	78:7	generations	97:14 249:12	goal 268:22
function 17:2	gathering 20:15	341:14	347:20	301:6 349:10
207:4 254:6,17	20:20 47:3	gentleman 9:11	glad 88:5	goals 86:20
259:2	71:10,13 72:16	gently 203:20	global 131:18	114:17
functionally	75:1 180:6	genuinely	206:19	god 355:3,5
14:3	381:22	259:15	globally 200:15	goes 10:11 21:19
functioning	gathers 22:2	geogretown 3:4	globe 154:1	91:16 155:20
245:21	gay 122:19	george 2:15 18:3	glove 322:19	171:12 181:2
functions	123:13	georgetown	go 8:10 17:9	201:19 211:9
104:22 207:20	geiger 3:19	1:12 4:7 117:5	22:10,13 28:22	211:16 278:16
217:14 252:4	295:14,14,20	georgia 3:3	43:9 48:20	351:4 356:14
254:3 273:5	326:8,12 327:2	110:14 143:10	51:2 54:5	going 10:4 23:7
298:8	331:19 340:8,9	german 343:17	57:12,18 67:19	23:22 24:7,21
fundamental	358:15 368:16	germany 343:20	75:20 79:19	24:22 25:1
132:2 226:21	371:3 375:16	getting 44:4	87:15,17,19,22	41:16,22 63:19
272:18 287:18	377:13,19	45:1 52:8	88:8 103:8	66:10 74:10
334:8	general 29:1	82:18 110:22	105:10 109:15	75:22 77:19,20
fundamentally	30:10 42:4	143:1 184:21	109:17 141:16	80:17 84:9,21
69:7 232:22	95:4 96:3	187:10 347:10	142:8 146:21	92:20 101:3
234:2 297:22	112:10 209:11	gewercman 7:16	148:18,19	102:4,4,17
further 34:21	209:12 215:13	giant 343:13	149:2 152:18	120:16 138:11
35:15 86:20	215:17 217:19	give 9:11 33:14	159:17 168:9	138:12,13,14
91:15 127:4	238:3,7,15	58:13 71:5,6	168:19 173:3	139:17 140:7
132:15 225:20	239:12 240:1	105:16 120:14	176:15 178:7	141:3,4 145:8
266:10 282:13	243:15,15	143:13 154:19	181:8 184:12	145:9,21,22
354:3 362:10	244:7,7 257:17	170:8 180:18	184:16 185:4,6	149:18 156:17
385:9	257:19 290:16	229:13 274:21	186:21 196:13	164:21 165:1
future 57:17	generalist	308:5 333:16	201:21 202:2,6	165:15 166:16
128:20 138:8	248:12	333:17 335:2	202:8 207:5	169:13,13
143:6 181:18	generalized 84:8	344:17 364:7	220:21 223:7	174:2,8 177:11
221:22 227:4	generally 19:18	given 5:14 23:17	238:1 245:17	177:16 178:21
233:15 262:22	69:6 90:10	90:7 111:11	261:3 262:13	182:20 183:6
293:21	238:15 265:15	114:22 119:10	263:17 269:20	184:19 185:11
	299:12 360:2	120:2 179:4	274:1 278:10	186:3 187:18
<u>G</u>	generals 295:6	205:20 233:4,4	285:6 286:17	189:14,18,21
gained 288:8	generated	246:6 250:7	286:20,22	189:22 191:14
gains 27:2	182:18	253:7 265:7	287:19 298:9	193:19 195:22
gap 156:14	generating	273:1 274:11	312:8 320:18	197:8 199:21
324:17	139:18 142:5	297:10 319:17	331:18 335:9	200:4 202:11
gather 72:22	182:17	334:8 358:11	346:19 347:18	207:4 212:5
74:20 105:1	generation	373:16 383:16	350:6 353:16	215:2 233:20
271:11,15	156:14 157:14	gives 21:12	357:15 372:11	234:18 241:16
	ı	1	1	

-				۷(
245 4 250 4 4		10 17 01 11 0	200 1 6 202 0	100 4 15 100 0
245:1 250:1,4	60:21 85:19	43:15,21 44:8	200:16 202:9	132:4,15 133:3
251:5,8,11	88:12 100:6	45:9 47:3 53:8	202:10,16,17	188:12 200:8
252:2 253:5	107:19 109:9	54:21 55:16	202:18 203:2,7	258:14 293:9
255:7 257:10	109:10 110:19	56:16,18 58:1	204:1,20	298:8 317:4
258:14 259:13	113:17,17	59:17 61:2	206:14 216:11	318:4 380:6
259:18 260:9	114:8 115:12	63:1,3 65:1	230:17 231:17	governmentwi
260:10,13,17	116:22 117:9	69:8,19 70:2	233:13 241:13	209:16
261:6,20	129:20 155:8	70:10 72:15	258:15,17,19	gps 15:7
263:11,13	157:20 158:2	73:3 74:19	259:1 263:4	grabbed 93:10
265:21 269:13	159:11,21	75:7,8,14,18	264:20 266:22	gradations
270:2,9,16	160:3,5,9	77:21 78:8,11	267:18 272:2	308:1
272:7 276:14	162:3 165:20	78:17,20 79:19	273:7 285:17	grade 283:14
279:7,9 283:5	176:2 177:6	79:20,22 81:9	294:14,22	grand 54:7
283:19 286:21	182:11 194:20	81:20 85:5	295:18 296:10	grant 301:20
311:19,19,20	197:6 198:6	88:21 90:2	297:12,18	303:9,9,15
312:1,10	203:4 207:5	93:3,15 94:21	298:18 299:17	318:19 319:21
316:12 319:8	245:22 256:20	95:12 102:11	299:19 300:2,8	338:10,11
322:3 323:17	263:5 275:1	102:14 103:14	301:9 302:3,6	342:9,10 348:7
324:13 330:9	277:14,15	103:17 104:3	302:12,18	348:13 356:12
339:10 342:3	279:5 284:22	104:19,22	303:2 314:6,10	362:22 366:11
348:17 349:19	294:22 321:16	105:2 107:12	315:2,7 317:7	378:12
352:16 353:1	321:17 327:5	108:4,6,12	317:17,18,22	granted 55:20
355:11,22	360:15	111:18 116:11	318:1,2 322:15	221:15 322:18
356:6 357:7,8	google 43:16	117:13,14	325:12 327:20	323:5
357:13,14,20	77:19 78:14	118:1 122:11	334:16 337:6,7	granular 11:14
359:20 360:10	112:3 339:13	124:7 127:13	337:10,13,19	101:13 210:10
361:8 362:16	googles 108:7	127:15,19	337:20 338:6	307:16 309:1
363:19 370:7,8	gotten 37:15	128:5 129:5,11	339:14,15	342:19 348:10
371:22 372:1	322:22 337:17	129:16 130:6	341:21 342:8	grateful 313:13
378:21 379:13	gov 7:8,11 384:5	130:16 131:1,5	343:1 344:9	great 42:22 43:1
380:20,21	384:9	131:7,12,14,16	345:2,15	43:17 74:14
goitein 2:13	governing 221:1	150:13 153:16	360:17,22	75:8 105:15
10:6,9 42:13	government	157:2 167:16	361:4 366:15	116:17 143:5
44:2 51:1,3,19	3:10 6:3 8:21	167:20,22	368:7,8 376:10	155:14,15
53:3,5 63:22	11:1 13:15	175:12,14	379:19 380:15	160:12 180:8
64:19 78:3	14:8 15:6,18	176:4,4,6,15	382:15 383:9	234:12 238:9
80:18 81:14	15:20,22 16:2	176:18 178:1	governmental	239:7 250:4
82:19,22 87:9	20:14,19 21:21	178:15 179:1,6	12:6 85:3 86:5	313:8 334:14
97:9 102:10,17	21:21 22:2	179:10 180:14	94:17 97:7	344:13 368:4
103:10,22	23:5,6,12,19	181:1 186:10	107:11	369:5 380:12
golden 186:6	24:4,11 27:14	186:16,20	governmento	382:18
good 4:2 18:5	27:21 28:1	187:3 189:5	228:14	greater 47:10
27:10 40:5	29:22 30:7,18	191:7 192:9	governments	62:4,5 83:13
49:6 53:9,11	39:7 42:15,16	199:16 200:1	45:1 57:21	84:5 86:19
.,.,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,	57.7 12.15,10	1,,,10 200.1	.5.1 5 / .21	0.1.2 00.17

87:2 104:3	253:22 256:6	199:19 318:11	harlan 18:2	17:7 18:11
130:5 204:15	257:12 271:21	322:19 385:12	harley 3:19	43:11 69:1,2
211:8 212:1	273:10,14	handle 64:15	295:14,14	150:6 201:1
213:6,14	279:8 326:16	127:8	337:9 358:1,8	heard 15:13
237:15 290:17	343:8 344:3	handling 37:15	358:11	83:15 96:11,18
300:3 301:10	354:4 358:13	231:11	harm 16:5 49:12	97:4,12 130:14
302:4 303:6	guessed 305:8	hands 151:17	50:20 54:6	139:1,16 140:7
337:11 376:20	guessing 156:8	156:8	55:5 59:7	166:19 199:11
greatly 176:19	guidance 159:8	hanging 305:17	237:2	235:19 269:11
greece 166:7	162:11 216:3	happen 15:11	harmful 49:7	280:2 348:1
green 9:12	218:1 312:7,13	105:8 224:18	121:15,22	383:9
207:5	guide 125:17	251:17 270:21	123:19 382:8	hearing 4:9,16
grind 45:3	226:20 276:7	300:11 348:17	382:11	4:17 207:9,18
grocery 148:18	282:2 293:20	happened 80:10	harmless 36:11	355:6 384:11
ground 28:11	guided 379:17	122:13 184:12	harmony 135:8	384:15
31:19,20 32:19	guidelines	311:1	harms 16:22	hearings 210:9
250:8 376:3	217:19 364:4	happening 24:1	50:5,7 52:6	hearsay 99:10
groundbased	gun 70:3	80:10 139:13	58:10 76:15	heart 208:7
369:17	guy 161:7	140:11 144:10	92:20,21 291:7	hearts 346:10
grounds 311:5	310:21	144:13 153:20	331:17,17	heavily 33:2
332:6	guys 139:5	154:6 175:1	336:9,20,21	182:22 375:3
group 3:20 25:6	171:10 188:10	241:18 252:20	340:17 350:17	hedge 380:14
116:1 161:15	244:9	255:2	harness 234:18	held 1:12 14:12
313:5		happens 49:14	hasnt 58:7 74:4	14:18 33:2
groups 41:10	<u>H</u>	57:5 109:17	97:5	91:2 132:13
251:1 336:10	habit 259:7	119:1 166:12	hasten 32:4	267:9 318:1
growing 40:17	hack 130:17	260:11 265:16	hate 291:22	325:10
62:15 147:12	hackers 158:15	happy 59:3	havent 87:18	hell 33:14
185:15 226:10	187:5 188:10	75:20 237:18	90:10 238:19	help 41:11 85:8
grown 104:21	hacking 130:20	278:10 326:8	283:21 381:20	105:1,1 116:1
growth 142:15	131:2	hard 197:10	hawk 183:22	128:5 133:3
142:20 143:3	hadi 3:7 134:14	246:8 249:6	haystack 257:4	160:12 161:6
guarantee 107:1	134:17 142:2,9	266:16 268:7	257:6	180:4 195:22
107:3 138:9,9	154:17 164:19	270:3 310:11	head 323:17	200:16 226:11
141:2	169:8 182:2	316:2 323:14	324:16	237:7 249:18
guarantees 42:9	189:7 hadnt 347:12	324:18 347:13	headed 240:3	271:3 289:21
138:7	half 124:13	371:21 381:5	headon 297:11	313:16 326:2
guard 261:7	137:15,16	harder 140:10	heads 215:22	335:17 340:4
guarding 291:6	139:10 163:13	195:20 314:3	headway 154:8	342:13 358:19
guess 61:8	284:3 368:1	317:11	health 37:13	377:3 378:14
161:18 165:21	halfway 108:9	hardwired	180:7,8 270:3	382:5,10,17
167:18 186:21	hand 58:18	308:18	healthcare	helped 129:20
189:6 194:18	68:19 110:9	hardwiring	332:18 hoom 5:16 6:3	206:18
195:1 224:1	00.17 110.7	308:14	hear 5:16 6:3	helpful 11:10
L				

116:6 159:9	hindsight 123:5	home 14:22 42:2	houses 53:9	106:1 116:9
164:19 202:19	hintze 3:6	113:17 141:11	hr 127:7	144:21 145:5
232:17 248:9	124:11,15	195:4	huge 186:17	165:16 179:8
254:15 260:3	134:9 149:5,12	homeland 25:8	250:15,16	208:6 227:1
276:6 314:5	151:4 152:3	100:5 104:6	310:7,8 320:3	229:11,13
354:2 383:17	153:18 168:3	226:2 290:4	320:5 356:7	233:15 272:22
helping 147:22	168:22 171:17	303:12,13	374:18	313:15 318:18
162:11 251:17	172:5 174:12	homes 151:10	hugely 144:16	318:19 322:13
helps 261:14,17	179:22 180:13	151:18 341:4	human 15:4	326:7,10
263:7 379:7	180:22 181:4	honest 107:7	26:17 51:5,5,6	330:17 333:16
382:2,3	185:21 188:1	honestly 87:18	51:14,16,20	354:10
heres 267:17	198:18 201:6	342:14	56:18 78:9	idea 50:17 108:6
hes 124:12	hintzes 144:14	honing 182:18	96:21 98:17	113:18 147:18
287:3	hipaa 99:19	honor 259:9	132:3 153:10	162:3 239:3
hesitate 80:2	145:17 229:12	313:9	155:10 174:15	269:2,21 276:9
heuristics	332:21,22	honored 226:5	349:5 350:1	287:10
143:16	historical	hoops 320:19	351:22	ideas 17:4 21:4
hew 162:21	162:21 229:14	hope 9:19 41:4	humandriven	152:15 171:14
hey 46:18	historically	88:16 126:8	348:21	291:3
120:13 148:17	122:3 229:15	143:3 172:21	humanly 56:15	identical 355:12
176:12 243:7	233:18	193:9 199:11	hundred 43:3	identifiable 37:8
269:19 274:2	history 44:12	213:3 225:19	56:4 93:17	113:8 229:8
hide 18:11 19:6	58:3,8 137:18	284:9 343:3	139:20,20	identification
166:2,6	231:20 284:2	hopefully	140:2 141:9	46:4 48:22
hiding 18:8,15	326:6 354:3	306:15,20	159:13 198:15	161:1 237:6
high 29:16	hit 48:21 311:19	307:9 359:1	379:1	identified 3:10
70:12 106:2	hits 259:11	hoping 140:19	hundreds	6:5 203:22
119:16 134:16	hitting 118:4	hops 91:9,10	325:16	identifier 34:17
163:5 170:16	hobby 271:8	horizontal	hurdle 282:11	identifiers 34:11
193:7 196:1	hold 62:20	316:17	hurt 160:21	39:9,11,18
289:14 305:14	226:21 286:16	horrible 174:1	hurts 112:18	identify 9:3 10:4
306:1	318:10 321:9	horse 271:8	hypothesis 56:3	36:1 68:7 90:5
higher 336:13	334:6 342:12	host 69:19	hypothetical	163:22 182:8
336:16	342:12 346:14	224:10	55:15 173:19	197:6 214:3
highest 29:12	362:13	hosting 226:5		232:17 235:16
33:2 152:20	holder 37:5	hotel 1:13 4:7	I	236:3,17
highlight 96:10	holding 26:3	hotter 347:11	ic 30:3 278:18	270:18 382:7
highlighted	62:13 91:6	hour 181:11	iccpr 51:7	identifying 36:9
91:13	110:1 207:17	hours 139:20,21	iconic 323:11	39:14 49:3
highlights	287:10 340:14	140:2	ics 120:7	183:15 196:12
120:19	holdings 43:20	house 170:2	id 19:12 20:12	196:14 198:7
highly 183:17	holds 237:1	369:18	22:21 33:18	381:17
hill 87:21	286:4	housekeeping	76:17 79:8	identity 34:20
304:21	holistic 233:1	203:12	82:20 91:21	35:2,3 39:20
	I	<u> </u>	<u> </u>	I

				1
59:22 113:21	102:3,4 107:8	immediate	36:6 76:13	impose 29:16
216:14	107:18 108:11	352:18	89:2,17 152:15	72:21 81:9
ideological	118:18 134:22	immediately	153:5 185:11	95:7 239:16
10:22	135:1 137:12	266:8,8	370:19 371:2	325:2
ideology 44:10	140:10,19,22	impact 3:2 6:1	implicit 98:11	imposed 322:19
ignore 48:2	149:1 153:18	17:1 109:13	282:17	323:4 324:6
ignored 352:2	160:11 165:15	121:11 172:18	implies 219:9	346:16
ii 122:7,18 187:6	166:20 167:1,5	175:7 204:10	imply 98:7	imposing 69:9
ill 17:22 41:17	167:8 168:9	220:12 224:21	importance	impossible
49:18 58:13	171:5 174:1	236:18 240:19	11:15 115:22	55:20 77:18
67:18 74:12	182:4 189:15	289:18,19	180:8 214:16	86:18 164:3
76:5 79:6	190:11 193:9	290:5 291:19	242:15 279:13	330:5 373:16
102:5 106:16	197:2,8 201:11	296:14 332:5	350:9	impoverish 17:2
107:22 109:16	201:20 207:3	375:5 382:11	important 11:18	impression
119:10 121:18	220:7 231:13	impacted 107:4	12:12 20:22	17:18 200:3
134:9 142:8	237:18,18	266:22	28:14 38:17	imprison 104:1
175:22 176:1,3	241:11,14	impacts 30:18	49:21 68:1	improperly
185:4,5 187:20	242:12 247:10	118:21 119:8	69:13 71:11	49:13 94:5
189:8 198:8	253:5 255:7	204:21 205:6	73:5,10 96:10	improve 114:16
217:7 250:10	259:5,21 262:6	266:13 291:8	96:13,14 100:1	165:9 200:10
257:18,18	263:10 269:6	291:17 336:9	120:5 124:17	224:20 240:18
276:11 279:6	270:16 271:21	336:19 381:17	136:15 143:6	357:2
279:11 280:2	277:14,15	381:18 382:8,8	147:3 153:5	improved
280:20 287:10	278:10 279:8	382:10	161:12,17,19	111:13 224:7
304:2 316:11	287:9 288:5	impediment	161:20 162:5	improvement
364:1 378:12	289:8,12	178:18	178:10 192:7,9	309:6
illegal 294:5	303:20 304:22	implement	196:4 207:9,21	improving
illegally 292:8,9	309:3,9 313:8	145:4 146:1	216:9,17	349:19
illuminative	318:6,16	206:19 342:13	224:21 227:20	impulse 85:19
74:2	326:17 328:22	implementation	239:4 240:8	inaccurate
illustrative	330:9 333:14	346:12	242:4 243:7	60:18
380:19	343:21 350:21	implemented	257:11 258:1	inaccurately
illustrious 10:4	350:21 351:14	84:13 92:12,16	259:2 260:20	294:17
im 10:4,19 11:12	357:12,14	100:10,11	264:1,16	inadequate 57:7
28:12 30:8	360:10 367:2	215:19 345:21	272:11,12,13	332:14
41:16,20,22	370:11 372:12	implementing	276:21 277:2	inadvertent
55:1 63:19	381:13	379:22	285:4 296:9	31:7 174:21
64:7 68:4	image 28:9	implements	312:5 313:14	inappropriate
69:21 73:8	68:10	348:4	329:22 331:15	98:7 265:20
75:20 80:19	imagine 87:2	implicates 121:5	335:12 336:6	350:18
81:15,15 82:11	306:12 378:3	implication	340:16,21	incentive 104:1
82:14,18 83:8	imagined 86:4	37:14 39:6	341:10 346:22	195:3
85:18 87:12	imbalance	107:20	359:4 372:10	inception
98:10,18 101:5	357:17	implications	374:7 379:5	119:11
	<u> </u>	<u> </u>	<u> </u>	I

				1
inching 369:10	290:22 309:1,2	326:15,22	inevitable	47:19 48:3,4
inchoate 266:20	incremental	327:10,19	178:13,20	48:13,14,19
incident 127:2	67:3	328:6,14 329:9	179:1,3,10	64:1,8 66:7,9
220:16	inculcating 95:1	329:18 330:2	200:1	66:22 67:3
incidental	independence	331:8,10,14	inexorability	68:6 69:8 70:3
262:20 279:3	4:22	334:11 336:15	142:11	71:10 72:16,22
incidents 316:14	independent	337:14 340:19	inexorable	74:20 75:11,15
include 21:2	29:17 31:1	364:22	142:4	76:14 78:7,8
47:3 115:19	41:10 57:20	individualized	infer 34:3 35:6,9	78:15,16,17
231:15 234:8	59:9 60:3,10	84:2	37:1 59:8	79:9,10,16,22
324:3	73:2 92:2	individuals 8:15	79:15,16	81:19,20 84:3
included 51:16	315:14 325:21	20:7 83:12	inference 35:14	87:16 89:7,15
141:12	373:22 374:1,2	84:6 91:12	37:3,8	90:6,16 91:2
includes 126:21	374:9,16	121:2 204:21	inferences 35:13	91:19 93:9,10
127:2,9 302:6	index 110:7	205:16 253:8	35:16	93:21 97:14,16
327:14	indiana 3:18	281:22 291:19	inferred 35:11	98:1,4 99:3
including 13:10	287:3	298:12 299:5	40:21 152:9	100:2,13
61:1 112:6	indicated 364:2	302:19 327:7	inferring 35:18	102:21 103:8
125:22 165:6	indirect 115:11	333:2 336:10	infiltrate 112:1	103:15 106:6,9
206:20 216:5	indirectly 183:4	336:10 337:5	infiltrations	112:6 115:7,10
312:17 370:6	indispensable	337:10 340:12	16:18	118:2,15 120:9
incorporate	296:19 333:6	340:15 352:10	infinite 308:5	120:22 121:2
234:19 310:1	individual 8:1	363:15 371:14	influence 115:4	122:8,10
incorporated	9:4 19:14,16	individuated	212:13 242:13	125:21 129:6
144:14	19:19 20:1,2,7	60:16	inform 5:18	129:13 131:5
incorrectly	20:8 44:7 49:2	industrial 20:18	226:11 261:14	133:5 134:6
73:11	49:7 54:15	industries 104:7	261:17 278:18	139:12 141:8
increase 50:15	59:5 60:8,15	industry 59:20	365:4	141:12 142:7
128:21 130:10	66:19 80:22	60:4,5 63:1,15	informant 13:16	142:16,20
141:3 169:15	81:3 82:3,8,15	104:21 117:13	information	143:20,22
increased 114:6	86:3 94:21	117:14 118:1	5:11 12:9,13	144:1 146:10
149:14 169:18	95:17 112:19	118:11 131:11	13:4,7,10 14:1	150:16 151:1,2
198:15	125:9 130:7	144:4 158:1	14:7,9,14,18	151:9 156:5
increasing 61:12	163:7 164:13	178:5,21 179:3	15:3,10,12	160:19 161:4
104:9 131:9	183:14 204:4	179:6 180:15	16:3 17:11,12	163:16 166:6
186:10,11	221:5,8 230:1	189:11 190:5	19:8 20:15	168:7 169:10
245:11	230:21 235:8	200:17 202:5,9	22:3,4,5 23:18	169:18 175:14
increasingly	236:11 268:14	383:10	24:20 33:14	176:12 186:8
104:20 108:5	270:15 271:18	ineffective	34:6,8,20 35:1	186:18 194:10
151:8 188:11	272:6,7 273:6	208:19	35:2,11 36:10	196:12,14
incredible	274:9 277:11	ineluctably	37:1,4 39:5,14	200:2 201:20
351:20	289:17 295:4	69:18	40:15,16 42:21	202:10 204:9
incredibly 17:10	295:10 296:21	inevitability	43:13,21 45:5	204:12,16
72:2 246:7	302:16,17	142:11	45:7 47:14,16	208:15 209:22
	I	I	I	

219:5,6 220:18 informing 145:7 211:9 239:5 insufficiently 222:14,19 infrastructure 325:9 334:16 332:9 332:0 332:3 312:4,6 342:18 insight 247:5 insight 320:12 insight 320:13 intellectually interlotions 38:6 intellectually interlotions 38:6 intellectually 27:5 interlotions 38:6 interlotion					J.
221:2 222:4,11 informs 208:14 222:10 244:1 299:12 content on the content of the conten	210.5 6 220.18	informing 1/15:7	211:0 230:5	incufficiently	175.4 316.15
222:14,19 infrastructure 111:15 310:9 339:3 insiders 115:3,6 intellectual interactions 38:6 interact 98:6 interact 98:6 interact 98:6 interacting 139:13 235:22 236:7,8 235:22 236:7,8 235:22 236:7,8 236:12,13 236:12,13 236:12,13 232:17 324:1 332:17 324:1 249:9,12 253:18,19 364:5 380:2 insiders 11:10 364:5 380:2 insiders 11:10 74:18 5:12 70:16 interactions 139:13 interaction 139:13 139:13 139:13 139:13 139:13 139:13 139:13 139:13 139:13 139:13 139:13 139:13 139:13 139:13 139:13 1		O		•	
228:22,22	ll '				
229:8 231:11 231:24,6 342:18 insiders 115:3,6 intellectually 275:9 interact 98:6 interacting 275:9 intellegence 275:9 intellegence 275:9 intellegence 275:4 intellegence 275:4 intellegence 275:4 intellegence 275:4 intellegence 275:4 intellegence 275:6 2	II '				- '
231:20 232:3 ingenuity 143:4 insight 247:5 335:3 interacting 139:13 335:45,15,17 331:8 322:13 338:8 322:13 338:52 236:7,8 332:213 338:322:13 insight 320:12 insight 247:55 insight 427:55 insight 427:55 insight 247:55 insight 247:55 insight 247:55 insight 247:55 insight 247:55 insight 247:55 insight 2	ll '				
235:4,5,15,17 235:22 236:7,8 236:12,13 322:17 324:1 249:9,12 253:18,19 265:2,4,6,8,20 266:6,12,17,19 267:1,3,7 268:10,14,19 268:21 280:4,8 281:18 283:10 296:8 299:1 300:9 302:14 300:13 307:7,8 307:13 310:21 337:11 205:19 300:9 307:13 310:21 337:13 30:21,17 30:13 30:4,11 337:13 30:4,11 337:13 30:4,11 337:13 30:4,11 338:16 235:7 318:10 10:10		′	,	•	
235:22 236:7,8 313:8 322:13 insights 320:12 215:4 intelligence 3:12 241:12 242:2 333:20 334:4 114:10 74:18 85:12 70:16 interactions		C •	U		0
236:12,13 322:17 324:1 241:12 242:2 333:20 334:4 14:10 15:13 30:4,11 70:16 interactions 15:13 30:4,11 70:16 interactions 70:16 interac				O	
241:12 242:2 333:20 334:4 114:10 15:13 30:4,11 74:185:12 70:16 134:19 249:9,12 351:6 360:13 253:18,19 364:5 380:2 16:5 380:2 16:5 380:2 16:5 380:2 16:5 380:2 16:5 104:9,13 266:6,12,17,19 267:1,3,7 61:5 104:9,13 268:10,14,19 268:21 280:4,8 27:21 339:22 281:18 283:10 266:8 299:1 26:20 279:17 300:9 302:14 initatives 240:8 300:15,16,19 303:3 307:7,8 218:7 232:13 307:13 310:21 323:9 251:2 331:10,15 331:10,15 331:10,15 331:10,15 331:10,15 333:7,12 338:7,8,21 247:6 339:27,7,22 inject 255:17,18 340:1,12,14 342:1,4 343:11 348:9 349:21 innocous 268:15 359:7,11,14,16 innocous 268:15 366:22 367:7 366:22 367:7 366:22 367:7 370:17 371:14 377:1 237:9 informative 383:8 inside 94:16 instrumenting 174:18 45:21 49:18	II '			- '	
246:12 248:6 249:9,12 351:6 360:13 27:1,5 107:10 90:18,22 18:11 119:15 110:13 254:10 261:4 inperency 55:11 266:6,12,17,19 267:1,3,7 61:5 104:9,13 268:10,14,19 133:16 235:7 268:12 1280:4,8 279:21 339:22 281:18 283:10 266:6,299:1 266:20 279:17 300:9 302:14 initatives 240:8 300:15,16,19 303:3 307:7,8 218:7 232:13 307:13 310:21 325:9 251:2 initially 9:10 320:9,11,20 320:9,11,20 339:27,22 initially 9:10 339:27,72 initiatives 212:7 339:27,72 inject 255:17,18 340:1,12,14 342:1,4 343:11 348:9 349:21 innocond 268:15 359:7,11,14,16 innocond 268:15 366:22 367:7 366:22 367:7 366:22 367:7 366:22 367:1 370:17 371:14 370:17 371:14 370:17 371:14 370:17 371:14 370:17 371:14 370:17 371:14 370:17 371:14 370:17 371:14 370:17 371:14 370:17 371:14 370:17 371:14 370:17 371:14 383:8 inside 94:16 instrumenting 174:18 70:16 interactive inspections 11:10 74:18 15:12 70:16 interactive inspections 11:10 74:18 15:12 70:16 interactive inspections 58:17 118:11 119:15 interactive inspections 58:17 18:11 119:15 interactive inspections 58:17 18:31 119:15 interactive inspections 58:17 18:11 119:15 interactive inspections 58:17 18:11 119:15 interactive inspections 58:17 18:11 119:15 interactive inspections 58:17 18:10 19:7 207:13,15 207:12,29:2 207:12,21:11,14,21 229:2 229:12 221:11,14,21 229:2 221:11,14,21 229:2 221:11,14,21 227:16 228:3,5 36:21 229:9 221:11,14,21 227:16 228:3,5 36:21 229:9 11:22 228:12 229:19 226:19 227:16 228:3,5 36:21 229:9 13:22 228:12 229:19 228:19 229:19 22	II '				
249:9,12 351:6 360:13 364:5 380:2 inspection 58:17 10:13 interest 8:22 9:1 10:13 instance 3:22 20:10:12 20:11:11 20:11:10 20:11:11 20:11:10 20:11:11 20:11:10 20:11:11 20:11:11 20:11:10 20:11:11 20:11:11 20:11:11 20:11:11 20:11:11 20:11:11 20:11:11 20:11:11 20:11:11 20:11:11 20:11:11 20:11:11 20:11:11 20:11:11 20:11:11 20:11:				,	
253:18,19					
254:10 261:4 265:2,4,6,8,20 266:6,12,17,19 267:1,3,7 61:5 104:9,13 295:6 298:7,13,18,22 33:7 66:21 33:16 235:7 268:21 280:4,8 279:21 339:22 295:4 209:12 210:11 211:10 67:2 73:17 296:8 299:1 267:0,279:17 300:9 302:14 302:15,16,19 303:3 307:7,8 218:7 232:13 307:13 310:21 335:9 251:2 initially 9:10 320:9,11,20 320:9,11,20 320:9,11,20 333:7,8,21 338:7,8,21 339:27,22 339:27,22 339:27,22 339:27,22 339:27,322 339:27,22 339:27,322 339:27,322 339:27,322 339:27,322 339:27,322 339:27,322 339:27,323 300:9 366:18 366:22 367:7 368:2 370:15 370:17 371:14 377:1 237:9 370:17 371:14 377:1 270:17 370:17 371:14 377:1 370:17 371:14	, ,		,	<i>'</i>	
265:2,4,6,8,20	II '		-		
266:6,12,17,19		O	-		
267:1,3,7				· · · · · · · · · · · · · · · · · · ·	
268:10,14,19	·		-	′	
268:21 280:4,8 279:21 339:22 95:4 209:12 212:11,14,21 91:22 97:15,15	, ,	,		, , ,	
281:18 283:10 inherently 20:18 26:20 279:17 300:9 302:14 initatives 240:8 302:15,16,19 303:3 307:7,8 218:7 232:13 161:9,14 227:16 228:3,5 361:21 377:19,20 initiality 9:10 256:9 231:6 232:6 89:20 137:12 320:9,11,20 90:15 271:12 initatives 212:7 29:9 71:22 243:7 244:2,8 154:16 160:1 247:6 334:13 337:11 215:19 216:2 159:15 378:1 340:1,12,14 injury 104:2 256:5 instilute 66:17 348:9 349:21 incocus 47:6 370:17 371:14 377:1 368:2 370:15 370:17 371:14 377:1 277:9 inputs 143:14 informative informative informative informative inside 94:16 instrumenting 174:18 45:21 49:18 17			_		
296:8 299:1 26:20 279:17 55:21 59:2 214:19 216:22 155:15 211:17 300:9 302:14 initatives 240:8 60:2 94:22 219:17 221:22 229:3 256:22 218:7 232:13 161:9,14 227:16 228:3,5 361:21 235:9 251:2 164:18 255:22 228:12 229:19 interested 80:20 231:10,15 242:10,15 242:10,18 149:9 153:22 247:6 239:9 71:22 243:7 244:2,8 154:16 160:1 247:6				, ,	,
300:9 302:14 initatives 240:8 60:2 94:22 219:17 221:22 229:3 256:22 230:15,16,19 303:3 307:7,8 218:7 232:13 161:9,14 227:16 228:3,5 361:21 307:13 310:21 235:9 251:2 164:18 255:22 228:12 229:19 interested 80:20 317:19,20 initially 9:10 90:15 271:12 instances 28:15 242:10,18 149:9 153:22 321:10,15 initiatives 212:7 29:9 71:22 243:7 244:2,8 154:16 160:1 334:13 337:11 215:19 216:2 159:7 161:9 245:1 253:15 220:10 272:22 338:7,8,21 247:6 195:15 378:1 254:4 255:19 367:6,8 372:12 339:2,7,22 inject 255:17,18 instill 357:18 256:12 258:2 385:11 348:9 349:21 innocent 256:13 institute 66:17 267:8,10 88:15 107:21 358:2,4,12 innocous 268:15 institution 30:7 275:11 276:11 216:18 226:17 358:2,4,12 innocuous 47:6 institutions 61:1 298:5 301:5 239:18 268:8 366:22 367:7 368:2 370:15 370:17 371:14 377:1 237:9 26:20 54:2 intent 243:21 intent 243:21 interests 2:10 interests 2:10 interests 2:10 interest 243:21 intent 243:21		•			
302:15,16,19					
303:3 307:7,8 218:7 232:13 161:9,14 227:16 228:3,5 361:21 307:13 310:21 235:9 251:2 164:18 255:22 228:12 229:19 interested 80:20 317:19,20 initially 9:10 256:9 231:6 232:6 89:20 137:12 320:9,11,20 90:15 271:12 instances 28:15 242:10,18 149:9 153:22 321:10,15 initiatives 212:7 29:9 71:22 243:7 244:2,8 154:16 160:1 338:7,8,21 247:6 195:15 378:1 254:4 255:19 367:6,8 372:12 339:2,7,22 inject 255:17,18 instilled 304:17 266:15,19 385:11 340:1,12,14 injury 104:2 instilled 304:17 267:8,10 88:15 107:21 348:9 349:21 innocent 256:13 institution 30:7 275:11 276:11 216:18 226:17 359:7,11,14,16 innocuous 47:6 107:10 133:10 304:7 320:3 282:18 339:8 366:22 367:7 inoperable 146:8 intended 217:15 357:16 365:8 370:17 371:14 237:9 26:20 54:2 intended 217:15 3:2,10 5:21,22 informative<					
307:13 310:21 235:9 251:2 164:18 255:22 228:12 229:19 317:19,20 320:9,11,20 90:15 271:12 321:10,15 34:13 337:11 215:19 216:2 159:7 161:9 245:1 253:15 220:10 272:22 338:7,8,21 247:6 195:15 378:1 256:5 385:11 340:1,12,14 342:1,4 343:11 348:9 349:21 358:2,4,12 359:7,11,14,16 360:9 366:18 360:9 366:18 360:2 367:7 370:17 371:14 377:1 3					
317:19,20			,	,	
320:9,11,20 90:15 271:12 instances 28:15 242:10,18 149:9 153:22 321:10,15 initiatives 212:7 29:9 71:22 243:7 244:2,8 154:16 160:1 334:13 337:11 215:19 216:2 159:7 161:9 245:1 253:15 220:10 272:22 338:7,8,21 247:6 195:15 378:1 254:4 255:19 367:6,8 372:12 340:1,12,14 inject 255:17,18 instilled 304:17 266:12 258:2 385:11 348:9 349:21 innocent 256:13 institution 30:7 267:8,10 88:15 107:21 358:2,4,12 innocous 268:15 institutions 61:1 298:5 301:5 239:18 268:8 360:9 366:18 innovative 39:3 210:1 353:5 358:8 344:4 357:10 368:2 370:15 101:7 146:8 381:22 357:16 365:8 370:17 371:14 237:9 26:20 54:2 intent 243:21 6:4 8:1,14 9:4 383:8 inputs 143:14 97:1 98:20,20 intentional 9:8 12:12 21:1					
321:10,15 initiatives 212:7 29:9 71:22 243:7 244:2,8 154:16 160:1 334:13 337:11 215:19 216:2 159:7 161:9 245:1 253:15 220:10 272:22 338:7,8,21 247:6 195:15 378:1 254:4 255:19 367:6,8 372:12 340:1,12,14 inject 255:17,18 instill 357:18 256:12 258:2 385:11 342:1,4 343:11 256:5 institute 66:17 267:8,10 88:15 107:21 348:9 349:21 innocent 256:13 institution 30:7 275:11 276:11 216:18 226:17 358:2,4,12 innocous 268:15 institutions 61:1 298:5 301:5 239:18 268:8 360:9 366:18 innovative 39:3 210:1 353:5 358:8 344:4 357:10 368:2 370:15 101:7 146:8 intend 313:20 interests 2:10 377:1 237:9 26:20 54:2 intent 243:21 6:4 8:1,14 9:4 377:1 237:9 26:20 54:2 intentional 9:8 12:12 21:1 383:8 inside 94:16 instrumenting 174:18 45:21 49:18	·	v			
334:13 337:11 215:19 216:2 159:7 161:9 245:1 253:15 220:10 272:22 338:7,8,21 247:6 195:15 378:1 254:4 255:19 367:6,8 372:12 339:2,7,22 inject 255:17,18 instill 357:18 256:12 258:2 385:11 340:1,12,14 injury 104:2 256:5 instilled 304:17 266:15,19 interesting 11:8 348:9 349:21 innocent 256:13 institution 30:7 275:11 276:11 216:18 226:17 358:2,4,12 innocous 268:15 institutions 61:1 298:5 301:5 239:18 268:8 360:9 366:18 innovative 39:3 210:1 353:5 358:8 344:4 357:10 368:2 370:15 101:7 146:8 intend 313:20 357:16 365:8 370:17 371:14 237:9 26:20 54:2 intend 217:15 3:2,10 5:21,22 377:1 237:9 26:20 54:2 intent 243:21 6:4 8:1,14 9:4 383:8 inside 94:16 instrumenting 174:18 45:21 49:18	ii ' '			′	
338:7,8,21 247:6 195:15 378:1 254:4 255:19 367:6,8 372:12 339:2,7,22 340:1,12,14 injury 104:2 instill 357:18 256:12 258:2 385:11 342:1,4 343:11 256:5 instilled 304:17 266:15,19 interesting 11:8 348:9 349:21 innocent 256:13 institution 30:7 275:11 276:11 216:18 226:17 359:7,11,14,16 innocuous 47:6 innocuous 47:6 107:10 133:10 304:7 320:3 282:18 339:8 366:22 367:7 inoperable instrument 381:22 357:16 365:8 370:17 371:14 input 211:11 instrumental 36:20 54:2 intend 313:20 interests 2:10 377:1 237:9 26:20 54:2 intent 243:21 6:4 8:1,14 9:4 383:8 inside 94:16 instrumenting 174:18 45:21 49:18					
339:2,7,22 inject 255:17,18 instill 357:18 256:12 258:2 385:11 340:1,12,14 injury 104:2 256:5 institute 66:17 266:15,19 88:15 107:21 348:9 349:21 innocent 256:13 institution 30:7 275:11 276:11 216:18 226:17 358:2,4,12 innocous 268:15 institutions 61:1 298:5 301:5 239:18 268:8 359:7,11,14,16 innocuous 47:6 innocuous 47:6 107:10 133:10 304:7 320:3 282:18 339:8 366:22 367:7 inoperable instrument 381:22 357:16 365:8 370:17 371:14 input 211:11 instrumental 381:22 357:16 365:8 377:1 237:9 26:20 54:2 intended 217:15 6:4 8:1,14 9:4 informative inputs 143:14 97:1 98:20,20 intentional 9:8 12:12 21:1 383:8 inside 94:16 instrumenting 174:18 45:21 49:18					
340:1,12,14 342:1,4 343:11injury 104:2 256:5instilled 304:17 institute 66:17266:15,19 267:8,10interesting 11:8 88:15 107:21348:9 349:21 358:2,4,12 359:7,11,14,16 360:9 366:18 366:22 367:7 370:17 371:14 377:1innocous 47:6 innocous 47:6institution 30:7 institutions 61:1 107:10 133:10275:11 276:11 298:5 301:5216:18 226:17 239:18 268:8304:7 320:3 107:10 133:10282:18 339:8 344:4 357:10368:2 370:15 370:17 371:14 377:1101:7 input 211:11 237:9146:8 instrumental 26:20 54:2357:16 365:8 intend 313:20 intended 217:15 intent 243:21369:2 367:7 368:2 370:15101:7 input 211:11353:5 358:8 intend 313:20 intended 217:15 intent 243:21357:16 365:8 intended 217:15 intent 243:21377:1 383:8inputs 143:14 inside 94:1697:1 98:20,20 instrumentingintentional 174:189:8 12:12 21:1 45:21 49:18					
342:1,4 343:11 256:5 institute 66:17 267:8,10 88:15 107:21 348:9 349:21 innocent 256:13 institution 30:7 275:11 276:11 216:18 226:17 358:2,4,12 innocous 268:15 institutions 61:1 298:5 301:5 239:18 268:8 360:9 366:18 innovative 39:3 210:1 353:5 358:8 344:4 357:10 368:2 370:15 inoperable instrument 381:22 357:16 365:8 370:17 371:14 input 211:11 instrumental intend 313:20 3:2,10 5:21,22 377:1 237:9 26:20 54:2 intent 243:21 6:4 8:1,14 9:4 informative inputs 143:14 97:1 98:20,20 intentional 9:8 12:12 21:1 383:8 inside 94:16 instrumenting 174:18 45:21 49:18	· ·	•			
348:9 349:21 358:2,4,12 359:7,11,14,16 360:9 366:18 368:2 370:15innocent 256:13 innocuous 47:6 innocuous 47:6institution 30:7 institutions 61:1 107:10 133:10275:11 276:11 298:5 301:5 304:7 320:3 304:7 320:3 353:5 358:8239:18 268:8 282:18 339:8366:22 367:7 368:2 370:15inoperable 101:7instrument 146:8381:22 intend 313:20 intended 217:15 intended 217:15 intended 217:15357:16 365:8 intended 217:15 intended 217:15 intended 217:15 intended 217:15377:1 informative 383:8inputs 143:14 inside 94:1697:1 98:20,20 instrumentingintentional intentional 174:189:8 12:12 21:1 45:21 49:18		9 6		′	
358:2,4,12 innocous 268:15 institutions 61:1 298:5 301:5 239:18 268:8 359:7,11,14,16 innocuous 47:6 107:10 133:10 304:7 320:3 282:18 339:8 360:9 366:18 innovative 39:3 210:1 353:5 358:8 344:4 357:10 368:2 370:15 101:7 146:8 intend 313:20 interests 2:10 370:17 371:14 237:9 26:20 54:2 intent 243:21 6:4 8:1,14 9:4 informative inputs 143:14 97:1 98:20,20 intentional 9:8 12:12 21:1 383:8 inside 94:16 instrumenting 174:18 45:21 49:18				,	
359:7,11,14,16 innocuous 47:6 107:10 133:10 304:7 320:3 282:18 339:8 360:9 366:18 innovative 39:3 210:1 353:5 358:8 344:4 357:10 368:2 370:15 inoperable 101:7 381:22 357:16 365:8 370:17 371:14 input 211:11 instrumental 313:20 intended 217:15 377:1 237:9 26:20 54:2 intent 243:21 6:4 8:1,14 9:4 informative inside 94:16 instrumenting 174:18 9:8 12:12 21:1					
360:9 366:18 innovative 39:3 210:1 353:5 358:8 344:4 357:10 366:22 367:7 inoperable instrument 381:22 357:16 365:8 368:2 370:15 101:7 146:8 intend 313:20 interests 2:10 370:17 371:14 input 211:11 237:9 26:20 54:2 intent 243:21 6:4 8:1,14 9:4 informative inside 94:16 instrumenting 174:18 9:8 12:12 21:1	′ ′				
366:22 367:7 inoperable instrument 381:22 357:16 365:8 368:2 370:15 101:7 146:8 intend 313:20 interests 2:10 370:17 371:14 input 211:11 237:9 26:20 54:2 intent 243:21 6:4 8:1,14 9:4 informative inside 94:16 97:1 98:20,20 intentional 9:8 12:12 21:1 174:18 45:21 49:18					
368:2 370:15 101:7 146:8 intend 313:20 interests 2:10 370:17 371:14 input 211:11 instrumental 3:2,10 5:21,22 377:1 237:9 26:20 54:2 intent 243:21 6:4 8:1,14 9:4 informative inside 94:16 97:1 98:20,20 intentional 9:8 12:12 21:1 174:18 45:21 49:18					
370:17 371:14 input 211:11 instrumental intended 217:15 3:2,10 5:21,22 377:1 237:9 26:20 54:2 intent 243:21 6:4 8:1,14 9:4 informative inputs 143:14 97:1 98:20,20 intentional 9:8 12:12 21:1 383:8 inside 94:16 instrumenting 174:18 45:21 49:18		_			
377:1 237:9 26:20 54:2 intent 243:21 6:4 8:1,14 9:4 informative inputs 143:14 97:1 98:20,20 intentional 9:8 12:12 21:1 383:8 inside 94:16 instrumenting 174:18 45:21 49:18					
informative inputs 143:14 97:1 98:20,20 intentional instrumenting 9:8 12:12 21:1 383:8 inside 94:16 instrumenting 174:18 45:21 49:18		_			·
383:8 inside 94:16 instrumenting 174:18 45:21 49:18					, and the second
		-			
mormed 00:5 1/0:15 19/:4 105:8 mtentionally 01:5 00:19			0		
	imormea 66:3	1/0:15 19/:4	103:8	intentionally	01:3 00:19

73:13,14,19	interpreted	223:10	116:4 246:18	iteration 271:16
79:4 97:3	301:15 355:20	investigative	252:8,8 304:12	iterative 263:13
101:14,16,20	interpreting	219:12 223:8	362:7 367:10	277:18
109:12 121:6	352:20	281:15	368:14 378:16	itll 269:6 346:10
151:3 203:22	interrupt 259:4	investment	isolated 36:14	ive 10:14 13:9
204:4 211:2,3	interrupt 239.1	127:11	228:13 263:20	31:4 43:8
231:11,13	110:16	invitation	isolation 36:11	70:14 88:15
256:3 264:4,10	intersections	303:16	issue 21:18 49:8	124:21 144:6
264:11 285:15	135:14	invite 206:2	59:5,6 87:15	208:17 212:19
295:10 334:2	interspersed	invited 6:14	113:6 125:3	214:13 242:6
340:17 367:12	228:17	9:21 224:17	131:19 177:7	251:20 256:16
interface 349:20	intervening 54:5	inviting 10:12	179:18 193:5	261:9 273:11
internal 30:2,10	55:4	207:8 216:8	193:18 196:8	274:6 288:8
57:21 95:20	intervention	295:22	229:9 249:3	290:3 293:13
173:1 193:2	60:10,10 77:13	involve 122:5	287:18 309:18	357:15 367:21
224:16 244:17	interview	200:21 201:3	320:5,13	ivy 197:4
			'	IVy 197:4
246:10 247:9	329:20	215:3 218:22	329:22,22	J
300:9 341:9	intimate 151:16	273:6 279:17	334:22 357:1	jackpot 278:14
346:13,14	introduce	308:19,20	378:16,17	jail 44:18 58:22
365:9 366:22	109:16 201:17	involved 58:11	issued 162:22	james 2:6 4:14
367:7 372:22	235:11	116:8 145:2,6	217:18,21	january 132:7
372:22	introduced	146:5 184:13	issues 5:19	january 132.7 japanese 122:7
internally 268:2	137:5	199:16 362:1	10:15 30:18	122:16
280:2,16	intruder 113:18	374:13	63:11 67:22	· -
338:20	intrusion 5:6	involvement	75:4,19 100:17	jargon 290:21
international	29:15 58:19,20	10:15	107:5 110:16	jerusalem
51:10 113:2	intrusions	involves 18:19	124:20 134:12	140:19
132:16 154:3,8	112:14	18:20,21 49:1	135:16 150:4,6	jim 96:6 109:6
206:18	intrusive 218:10	122:1 279:20	153:8 154:4	154:2 159:19
internet 43:17	218:16 219:3	iot 139:1	171:21 172:14	164:20,21,22
115:1 131:2	221:17 261:2	iphone 137:4	173:4 176:20	278:3
139:1 143:11	262:17 341:2	ireland 132:14	205:2 224:16	jims 107:18
143:12 160:17	364:10	ironic 337:16	225:15 240:7	job 12:10 27:10
185:17 186:5	inured 147:18	351:5	247:2,18,18	42:22 43:1
335:21	invades 120:21	ironically	248:22 276:4	59:19 95:13
internment	invest 188:16	361:14	284:19 285:7	156:21 171:14
122:17	312:6 347:7	irreconcilable	289:17 293:22	177:6 224:18
interpretation	invested 316:20	55:8	295:17 352:15	226:17 240:15
301:11 314:21	investigation	irrelevant	353:14 379:7	243:12,16,18
interpretations	218:7 219:10	350:11,21	issuing 29:18	244:12 267:5
71:17 315:1	221:21 352:7,9	irs 69:10	item 36:7,8,20	280:14 283:2
325:19	investigations	isnt 18:18 19:22	36:21 201:15	304:18 305:5
interpretative	217:21 219:8	23:14 49:10	items 34:2 36:15	319:15 324:21
356:9	219:14 221:9	54:15 75:8	91:11	325:6 349:15
	I	I	l .	I

				1
357:15	270:20 274:22	147:15,17	176:15 178:17	148:17,20,21
jobs 244:4	judicary 254:11	keep 7:1 19:1	187:7 212:9	149:1,13 152:3
joe 6:22 7:16	judicial 29:12	27:17,18 28:12	214:9 261:11	152:7,11
joel 3:11 206:4,8	42:7 60:9	76:1 94:3	261:12 280:4	153:12,19
207:1,7 241:11	70:22 73:3	102:7 105:5	325:19 328:12	154:15,17
241:19 247:15	85:21 106:22	109:20 113:17	339:4 346:4	155:22 157:5
253:15 257:18	130:9,12	133:1 151:9	347:1 361:3	157:13,14,19
257:21 258:8	168:13 221:13	162:4 200:1	362:4 369:8	158:3,10
259:6,9,12	282:8	255:6 282:14	372:16 373:2	160:11 161:11
260:1 262:6	judicialmade	306:6 322:8	373:20	163:11 168:9
264:14 275:22	369:3	323:16 359:11	kinds 45:15 46:2	170:1,2,3
279:5 310:5	judiciaries	362:16,19	64:10 73:17	172:5,6 175:4
john 18:2 303:9	32:20	keeping 18:19	90:21 103:15	176:8,18 180:8
315:9,10	judiciary 74:6	49:10 64:7	161:4,5 219:20	182:3,21
316:11 348:7	117:7 179:16	316:1 374:20	249:3 264:18	183:21 184:21
356:11 366:8	193:21 210:5	kelly 6:22 7:16	280:3,5 308:13	189:22,22
join 316:17	julian 66:16	kept 18:22,22	349:11	190:16 191:19
joined 242:19	jump 356:12	19:8 22:5	kinetic 298:12	193:13 196:1
joining 205:16	378:12	49:10 94:6	klatch 77:3	197:9,16,18
206:14 225:22	jumps 366:11	336:7 341:7	knew 17:21 77:1	201:7,14 209:5
joke 321:7	june 137:4	key 22:20 37:7	77:1 180:4	213:18,20
jonathan 8:3	jurisdictions	48:16 62:18	191:10 202:17	214:17 215:5
jones 76:8 369:7	220:20	74:8 83:11	356:6	217:12 218:4
journalists	jurisprudence	128:22 193:11	knocked 120:13	220:18 222:17
115:4	15:1 292:14	208:20 214:1	know 13:14 23:7	223:22 240:22
judge 7:19 10:9	jury 54:8	218:9 252:5	23:11 24:19	241:18,20,22
25:9 29:18	justice 2:13 3:13	265:12 266:3	28:6,6 43:2	242:8 247:7
80:18 179:12	133:13 209:15	kick 206:22	45:2 50:1,14	248:14 249:5
352:12 372:8	213:2 215:12	kid 196:1	56:21 60:15	251:6,7,9,18
373:4,15 374:5	222:7 246:7	kind 10:11	62:9 67:11	254:14 255:16
377:11	364:15 372:15	13:20 15:17	69:21 70:4,13	256:16 257:12
judged 54:21	justices 225:5	20:4 22:1 27:6	82:12 83:6	257:15 259:18
284:9	justification	30:16 46:1	87:14,20 88:2	260:1,17
judgement	63:12 72:22	50:22 54:9	88:12 93:12	261:10,15
25:15	justified 260:16	57:22 62:9	95:1,11 101:18	262:16 263:6
judgements	366:2	64:6,14 68:10	103:6,7,19	263:19 267:11
32:19	justify 22:8 63:5	68:14,18 72:7	104:14,16	268:5 269:3
judges 352:12	63:6 106:10,11	81:11 85:14	105:15 106:2	270:6 271:7
360:20 373:11	315:4 317:7	86:18 91:13,14	108:5,10 120:8	275:18,18
373:13,18	justin 368:3	101:13,18,19	120:15 121:4	276:17 277:19
377:20	K	108:7 145:1	122:13 138:7	278:14 279:18
judgment 26:19		152:17 159:6	139:17 142:21	280:16 281:11
27:1 33:6	kaplan 286:3 katz 74:16 75:17	159:13 160:22	143:17 144:6	281:18 282:4
214:7 257:8	Katz /4.10 /3.1/	165:7,7 176:12	147:16 148:7	286:5 289:21

				<u> </u>
200 2 201 2 4		0.10.10.2.4	220 10 201 11	
290:2 291:2,6	labs 144:4	8:19 18:3,4	230:18 301:11	learning 35:10
292:5 293:3,17	lack 48:11 56:3	22:8 26:18	302:2 321:4	191:1,8 205:14
296:22 304:1	127:16,20	42:12 51:10	341:7 343:15	251:20
306:18,19	193:8 253:3	65:9 69:10	lawsuits 129:10	leave 156:8
309:19,20,21	291:4 310:3	70:16 72:6	lawyer 80:3	186:15 187:8
312:1 314:18	344:8	97:12 112:6,18	135:1,2 247:22	187:12 224:2
320:15 322:5,8	lacked 122:3	117:5,6,8	305:8	319:13
322:20 325:10	lacking 327:11	124:22 131:6	lawyers 165:2	leaves 277:4
329:14,18	lacks 179:17	132:10,22	187:12 249:6	leaving 256:4
330:14 332:18	248:15	133:20 145:3	305:11 309:19	led 274:5 290:2
332:20 337:15	lag 121:18 122:5	145:11 146:11	311:13,15	lee 3:13 215:10
339:18 342:17	laid 46:11 128:4	147:4 168:4	354:14 378:3	215:17 216:7
344:2 345:4,14	landscape	169:1 177:10	layer 192:12	239:17 245:22
348:14 350:14	227:21	177:14 189:16	layering 82:2	263:11 267:13
351:2 352:6,7	language 222:18	191:6 217:3	layers 209:9	277:16 281:9
352:19,20	249:6 375:11	219:17 220:18	212:22 223:21	left 4:22 9:20
353:13 354:9	languages	221:7 223:4	lazy 187:8	187:18 286:6
355:16 360:14	158:18	224:14 243:20	lead 38:2 47:19	347:6 370:2
361:2,7,13	languish 359:14	247:18 248:1	164:2 217:7	378:19
364:13 367:16	lapd 182:7	267:15 272:16	328:20 330:7	leftover 259:8
368:1 369:8	large 37:11 40:2	282:2 287:3	331:21 335:15	leg 345:9
371:13,19	40:17 50:17	292:5 298:22	leadership	legal 13:19 21:9
374:11 376:4,5	58:20 62:1	304:6 309:20	238:13	41:3,7 42:7
377:19 378:4	66:20 67:5,6	311:14,21	leading 110:16	63:16 80:3,9
378:16,18	117:17 125:10	315:1,5,11,18	206:4	98:14 129:11
379:6 380:11	171:19 178:4	327:8 341:8	leads 47:13	132:16 150:4
381:19 382:22	211:10 246:6	343:6 352:6	206:10	174:3,5 191:7
knowing 58:19	248:2 302:18	355:1,18,19	leagues 197:5	216:3 230:4
227:6 241:16	333:10 371:13	356:5,8 364:6	leak 115:9	238:16,17
251:7 272:22	largely 235:1	364:8,12 365:1	leaked 49:13	247:18 272:18
338:6	289:10 307:9	369:14	leaking 115:3	292:4 293:7,14
knowledge	315:8 380:2	lawabiding	leaks 198:12	293:19 316:18
47:19 50:18,19	larger 61:2,6,6	133:10	leaning 33:1	334:14 348:6
105:21 176:6	83:20 249:19	lawful 189:6	leaping 101:4	348:12 354:11
288:9	330:3	260:15,16	learn 47:8	355:2 377:7
known 36:3	lastly 140:22	261:21 262:16	147:22 149:16	legality 187:13
76:6 138:2	191:11 333:4	265:6	190:22 191:18	legally 8:6 68:3
166:9	latanya 195:12	lawfully 209:1	197:21 311:13	243:2
knows 77:11	latent 117:19	laws 9:6 24:10	311:14 334:22	legislation
156:19 306:21	latest 40:9 112:3	48:17 103:19	341:22 342:8	303:19 378:19
337:10 380:20	latitude 51:13	129:14 144:20	learned 3:17 6:6	legislative 212:7
	laud 255:16	145:7 146:8,14	10:14 92:3	legislators 8:21
L	299:10	168:22 178:2	230:10 244:9	legislature
label 152:12	law 2:15 3:18	216:4 230:16	285:5 304:19	192:21 193:4

legitimate 9:7	203:5 205:10	likehealth	256:15 259:14	347:5,14
86:12 147:7	206:5,11	235:22	326:1 341:15	live 20:4 158:12
231:10 292:22	207:16 209:2	likelihood 31:7	347:18	191:15 205:14
legitimately	209:13,19	likeness 5:10	lines 132:10	245:17
361:21	211:3 215:11	likes 140:19	162:16 173:7	livelihood 59:11
length 286:22	215:14,18,21	limit 40:14,16	270:7	lives 12:4 34:6
lens 33:15	216:1 217:17	68:6 90:9	link 15:18 34:12	49:5 151:16
216:22 223:19	223:17,19	106:6 112:15	39:11,19 161:4	379:12
340:11	225:12,16,18	124:6 143:3,4	linked 91:20	living 126:5
lenses 263:7	226:6,9 227:2	143:22 178:10	linkedin 137:1	283:14 356:22
370:5	227:18 229:5	178:17 293:15	141:11 339:1	livingston 1:22
lesbian 122:19	229:14 231:1,3	293:19 299:1	linking 49:1	385:4,16
123:13	231:13,14,18	336:21 351:3	90:21	liza 2:13 10:6
lessons 3:17 6:6	231:22 232:4	351:11 372:10	links 35:3 90:5	41:22 70:8
36:16 58:4	232:18,19	limitation 329:7	90:22	74:15 103:22
92:5 344:4	233:5,10 234:5	350:13 371:9	list 10:18 83:14	108:2
letters 129:18	234:16 237:6,7	372:6	83:16 84:19	loaded 255:12
letting 160:2	237:12 238:10	limitations 28:1	120:14 288:12	local 220:19
370:17	238:12 240:3	226:12 297:7	listed 51:6,7	location 114:1
level 21:19	242:15,22	297:14 312:9	299:15	148:3 168:7
23:14 28:22	243:8,19	312:15 327:15	listen 166:10	232:6
29:12 30:8	244:10,13,18	342:7 350:8	listing 311:9	locations 39:16
86:14 93:14,19	275:7 280:22	360:3,5 362:13	lists 55:18	lock 194:20
108:13 127:9	282:22 284:13	362:20 372:21	literally 119:11	locked 306:15
144:9 163:10	289:20 295:17	375:21 379:18	353:6	306:15
164:12,13	295:21 296:3	382:17	literature 40:17	locks 113:16
175:2 183:12	303:10 304:18	limited 77:1	little 45:14	log 306:20
183:14 184:22	305:5 310:22	118:6 120:2	47:15 53:21	310:12
185:21 192:8	317:10 334:1	182:5 271:20	54:19 57:18	logging 308:10
193:13 210:10	349:13 351:18	306:19 319:8	60:2 83:1	320:20
213:2 253:9	353:8 356:21	343:2 352:22	92:18 110:7,10	logistically
257:17 270:8	357:7 358:21	373:7 376:5	114:21 146:1	285:22
273:21 305:15	383:20	limiting 39:2	159:3 162:19	logs 320:7,17,18
306:1,10	liberty 10:7 59:5	54:4 142:15	165:19 179:19	321:1,6
307:16 309:16	59:10 135:15	256:3 299:7	190:21 199:20	long 22:4,7 32:9
310:14 342:19	317:12	300:14 301:3	205:20 217:9	55:22 62:2
343:3 377:21	life 17:20 126:13	limits 22:3	218:18 220:9	94:3 118:2
leverage 249:13	324:10,11	38:14,17,19,21	227:1 229:13	123:9 126:2
250:4 251:14	lifetimes 341:5,6	38:22 141:18	233:16 248:10	138:16 202:6
levied 227:8	lift 170:11,19	292:17 330:16	252:2 262:15	229:6 284:2
lgbt 122:2 123:4	light 5:9 92:2	351:20	270:17 271:7	300:18 309:7
liberal 235:12	178:11 207:4	line 44:21 67:19	277:4 288:8	319:16 362:13
liberties 1:3 4:3	220:9 243:21	75:7,20 108:3	305:14,18	362:16
111:4 116:10	lights 22:14	120:8 159:17	323:22 330:12	longer 40:6 56:2

118:8 138:15	339:2 340:17	83:6,18 86:9	loyalty 37:5	malware 189:1
158:14 191:10	340:22 347:13	88:18 94:11	lunch 7:4	manage 139:19
282:19 289:1	363:19 364:2	95:3 96:11,18	lynn 7:15	140:4 237:2
362:19	366:16 368:14	120:9 123:8	286:15	307:18 308:6
longestablished	371:5,12 373:3	138:12,14,17	lynne 1:22 385:4	310:4,13
118:5	373:21,22	139:5,14,16	385:16	342:18 350:2
longstanding	382:9	143:8 144:19		management
332:12	looked 50:15	144:20 145:3	M	127:3 234:11
longterm 59:6	160:16 281:13	145:10,17	m 1:14 4:5	248:21 290:18
look 11:16 15:21	looking 68:11	147:4,6,7	machine 35:10	290:19,22
22:19 26:8	89:6 90:21	154:9 157:5	machines 349:2	291:3,13,21,22
30:14 41:13	105:6 141:1	176:18 183:1	magicmarker	304:4 305:19
43:14 48:7,8	142:10 196:19	187:2 191:12	307:12	307:15,16
50:3,13 52:13	197:11 220:2	191:16 192:22	mail 107:14,17	381:16 382:1
52:15 59:16	236:19 240:17	193:14,19	main 36:16	managing
61:18,20 66:18	240:18 252:9	197:13 212:1	184:5 242:5	124:19 306:2
75:3,10 76:9	263:6,18 269:2	245:10 247:1	305:10 371:8	310:7
76:11 78:6	269:22 270:15	251:19 252:12	mainstream	mandate 5:19
82:17 87:17,20	282:8 288:18	262:19 264:15	10:22 229:9	231:5
87:22 88:4	320:12 321:3	266:2,5 268:13	maintain 191:2	mandating
91:9 95:15	339:6 341:13	269:8,9 273:5	191:3 208:11	165:22
104:16 105:6	343:10 357:19	273:7 275:5,9	214:11 370:16	mandatory
116:22 120:10	366:6 367:3,4	278:7 283:13	maintained 57:6	172:12
120:16 141:15	368:9 371:18	284:6 290:6	226:22	manipulate
156:13 157:7	372:2	291:13 292:1	maintaining	228:20
170:4,5,5,6	looks 15:16,22	304:1 305:9	140:9 190:3	manipulated
177:1 178:15	73:3 236:16	307:15 308:3,9	214:17,17	336:15 378:11
178:16 197:2,3	loop 183:20	309:3,13	maintains	manner 103:18
198:22 199:1,3	loophole 300:12	317:16 320:8	265:12	298:4 378:8
215:3,6 223:10	loosen 327:6	328:4 333:1	major 241:20	manners 21:17
223:18 228:8	loosened 378:6	342:10,19	262:1 298:14	manuel 217:22
233:15 236:22	los 182:7	343:4,7 344:6	298:21 330:19	map 343:15
240:20 243:18	lose 97:14	344:13 348:8	majority 53:15	mapping 155:9
250:21 252:2	257:10 268:18	349:18,19	64:20 253:12	marcos 250:19
256:11 257:15	losing 95:13	368:2 370:1	makers 349:12	marginal 73:16
258:1 261:13	148:13 270:11	375:20 383:11	making 7:17	marked 222:2
261:19 269:7	loss 16:5	383:16	18:21,22 32:19	markers 242:3
269:19 270:17	lost 16:4 24:10	lotion 35:9 37:6	35:15 100:18	market 24:5
272:2 273:17	326:5	lots 86:1 103:19	140:13 142:2	42:21 63:16
275:3 276:15	lot 21:12,13 25:2	220:2 223:21	148:17 154:7	marketing
276:16 310:12	43:11 47:4,6	loudly 154:4	154:14 167:9	127:6 172:16
311:5 320:22	48:12,17 49:12	love 189:12	199:5 246:13	320:4
321:6 322:3,7	49:17 71:18,18	low 305:17	249:8 265:11	marriott 1:13
329:19 338:22	71:20,21 72:5	lowly 357:11	379:21	4:7 206:17
	I		ı	1

283:15 284:6 mathematically 355:20 356:1 276:3 281:5 360:10 363:1 357:22 369:19 371:3,8 369:19 371:3,8 371:21 377:11 377:22 377:22 377:22 377:22 377:21 377:22 377:22 377:22 377:22 377:22 377:22 377:22 377:22 377:21 377:22 377:21 377:11	<u></u>				
marriotts 144:9 149:20 265:20 268:1 99:5 74:12 101:6 206:19,20 151:5,8 157:17 286:7 289:6 medicine 304:9 109:9 110:4,5 marshall 18:2 160:7 161:20 295:6 299:7 medicine 304:9 109:9 110:4,5 maryland 385:5 180:17 184:22 332:21 339:10 medice 2: 3 4:2 123:4,13 mass 60:16 188:1,7,17,18 355:12 356:2,2 79:21 80:4,6 154:12 159:1: 301:7 310:15 190:16 191:17 356:3 360:6 80:13 107:2 181:9 185:7,8 masses 43:17 195:11 202:10 meant 180:5 108:19,22 252:16 254:8 256:10 massively 210:18 214:20 334:2 352:9 109:5 116:14 254:18 256:1 289:13 320:6 248:14 273:4 measure 119:8 165:15 166:18 286:14 295:2 335:20 289:10 290:20 345:18 169:8 17:9 378:21 383:6 matching 12:19 327:18 332:15 366:20 340:9 32:22 68:5 284:17,22 290:3 mathematics 355:20 356:1 23:22 268:5 284:17,22 356:11 357:22	247.21.264.8	103.13 121.20	237.4 252.1	medical 08·21	32.5 /11.17 10
206:19,20					
marshall 18:2 160:7 161:20 295:6 299:7 medieval 76:21 110:6 122:19 marvelous 26:2 168:3,22 169:3 332:21 339:10 medine 2:3 4:2 123:4,13 maryland 385:5 180:17 184:22 339:18 352:22 4:20 74:14 141:17,21 mass 60:16 188:1,7,17,18 355:12 356:2,2 79:21 80:4,6 154:12 159:1: 301:7 310:15 190:16 191:17 356:3 360:6 80:13 107:2 181:9 185:7,8 317:2 365:17 195:11 202:10 meant 180:5 108:19,22 252:16 254:8 massive 195:2,5 232:9 244:19 measure 119:8 116:18 165:14 254:18 256:1 massively 277:3 282:1 221:11,18 167:18 168:16 286:14 295:2 match 85:9 91:8 291:3 321:11 measured 194:6,14,18 383:21 383:6 matching 12:19 327:18 332:15 measures 23:21 264:2 268:9 290:3 matching 12:19 327:18 332:15 323:22 68:5 284:17,22 290:3 mathematically 367:10 368:17 276:3 281:5 360:10 363:1 125:11 12:2 113:9					
marvelous 26:2 maryland 385:5 168:3,22 169:3 332:21 339:10 medine 2:3 4:2 d. 20 74:14 d. 141:17,21 mass 60:16 l. 188:1,7,17,18 l. 355:12 356:2,2 mass 60:16 l. 190:16 191:17 l. 356:3 360:6 meant 180:5 mess 43:17 l. 195:11 202:10 meant 180:5 l. 108:19,22 l. 252:16 254:8 messive 195:2,5 l. 232:9 244:19 measure 119:8 l. 116:18 165:14 l. 254:18 256:1 massively l. 277:3 282:1 l. 21:11,18 l. 167:18 168:16 l. 286:14 295:20 master 180:5 l. 168:19,22 l. 254:18 256:1 l. 254:18 256:1 massively l. 277:3 282:1 l. 221:11,18 l. 167:18 168:16 l. 347:20 352:17 l. 335:20 l. 289:10 290:20 l. 345:18 l. 169:8 171:9 l. 378:21 383:6 match 85:9 91:8 l. 291:3 321:11 measured matching 12:19 math 235:11 l. 340:10 353:6 l. 23:22 68:5 l. 23:22 25:19 l. 36:11 357:22 l. 290:3 mentioned 70:5 methanics l. 377:21 l. 377:21 l. 377:22 mechanism 33:4 l. 377:22 l. 377:11 l. 377:22 mechanism 33:4 l. 377:22 l. 378:11 mechanics l. 377:21 l. 377:12 l. 378:21 383:4 l. 377:22 l. 263:22 319:18 l. 314:11 l. 182:5 190:17 mechanism 3:4 l. 190:19 l. 379:11 l. 35:22 247:4 l. 291:10 263:15 l. 171:1 193:15 l. 221:16 223:2 208:7 242:1 meaningful l. 253:22 234:4 l. 291:10 263:15 l. 379:11 l. 35:22 294:18 l. 145:19 190:19 l. 35:21 263:22 319:18 l. 145:19 190:19 l. 35:21 263:22 347:4 l. 263:18 282:2 l. 299:10 299:9 l. 56:8 57:7,13 megi 139:10 mentioning mellon 251:8 l. 223:20 l. 123:4,13 l. 141:17,21 l. 141:19:18 l. 154:12 l. 154:18 l. 141:17,21 l. 141:19:18 l. 141:19:18 l. 169:18 l. 140:4 l. 141:17,21 l. 141:19:18 l. 141:19:18 l. 141:19:18 l. 141:19:18 l. 141:19:18 l. 141:19:19 l. 141:19 l.	II '	,			
maryland 385:5 180:17 184:22 339:18 352:22 4:20 74:14 141:17,21 mass 60:16 188:1,7,17,18 355:12 356:2,2 79:21 80:4,6 154:12 159:1: 301:7 310:15 190:16 191:17 356:3 360:6 80:13 107:2 181:9 185:7,8 317:2 365:17 195:11 202:10 meant 180:5 108:19,22 252:16 254:8 massive 195:2,5 232:9 244:19 meanure 119:8 116:18 165:14 254:18 256:1 289:13 320:6 248:14 273:4 123:22 141:6 165:15 166:18 286:14 295:2 massively 277:3 282:1 221:11,18 167:18 168:16 347:20 352:1 335:20 289:10 290:20 345:18 169:8 171:9 378:21 383:6 match 85:9 91:8 291:3 321:11 measured 194:6,14,18 383:21 384:3 matching 12:19 336:20 340:9 measures 23:21 264:2 268:9 210:6 220:14 math 235:11 340:10 353:6 213:22 25:19 356:11 357:22 290:3 mathematically 367:10 368:17 276:3 281:5 360:10 363:1 125:1 142:22 mathematics </td <td></td> <td></td> <td></td> <td></td> <td></td>					
mass 60:16 188:1,7,17,18 355:12 356:2,2 79:21 80:4,6 154:12 159:1: 301:7 310:15 190:16 191:17 356:3 360:6 80:13 107:2 181:9 185:7,8 317:2 365:17 195:11 202:10 334:2 352:9 109:5 116:14 254:18 256:1 masses 43:17 210:18 214:20 334:2 352:9 109:5 116:14 254:18 256:1 massive 195:2,5 232:9 244:19 measure 119:8 116:18 165:14 278:5 286:10 289:13 320:6 248:14 273:4 123:22 141:6 165:15 166:18 286:14 295:20 massively 277:3 282:1 221:11,18 167:18 168:16 347:20 352:15 335:20 289:10 290:20 345:18 169:8 171:9 378:21 383:6 match 85:9 91:8 291:3 321:11 measured 194:6,14,18 383:21 384:3 matching 12:19 327:18 332:15 291:18,18 196:7 203:4 mention 147:1 270:1 283:13 354:13,18 112:8 163:22 354:6 355:7 290:3 284:15 284:6 355:20 356:1 213:22 225:19 356:11 357:22 99:22 115:21 mathematically		,			'
301:7 310:15 190:16 191:17 356:3 360:6 meant 180:5 108:19,22 252:16 254:8					· ·
317:2 365:17 195:11 202:10 meant 180:5 108:19,22 252:16 254:8 masses 43:17 210:18 214:20 334:2 352:9 109:5 116:14 254:18 256:1 massive 195:2,5 232:9 244:19 measure 119:8 116:18 165:14 278:5 286:10 289:13 320:6 248:14 273:4 123:22 141:6 165:15 166:18 286:14 295:26 massively 277:3 282:1 221:11,18 167:18 168:16 347:20 352:1 335:20 289:10 290:20 345:18 169:8 171:9 378:21 383:6 matching 12:19 327:18 332:15 321:18 194:6,14,18 383:21 384:3 matching 12:19 336:20 340:9 measures 23:21 264:2 268:9 210:6 220:14 math 235:11 340:10 353:6 223:22 68:5 284:17,22 290:3 283:15 284:6 355:20 356:1 213:22 225:19 356:11 357:22 99:22 115:21 mathematics 371:21 377:11 377:22 mechanicsm 383:3 384:13 144:21 153:7 80:4,5,8 90:21 98:12 149:2 322:3 263:15 16:12 193:16 22 229:16 22		, , ,	,	· · · · · · · · · · · · · · · · · · ·	
masses 43:17 210:18 214:20 334:2 352:9 109:5 116:14 254:18 256:1 massive 195:2,5 232:9 244:19 measure 119:8 116:18 165:14 278:5 286:10 289:13 320:6 248:14 273:4 123:22 141:6 165:15 166:18 286:14 295:20 massively 277:3 282:1 221:11,18 167:18 168:16 347:20 352:17 335:20 289:10 290:20 345:18 169:8 171:9 378:21 383:6 match 85:9 91:8 291:3 321:11 measured 194:6,14,18 383:21 384:3 matching 12:19 327:18 332:15 291:18,18 196:7 203:4 mention 147:1 material 324:15 336:20 340:9 measures 23:21 264:2 268:9 210:6 220:14 math 235:11 340:10 353:6 23:22 68:5 284:17,22 290:3 mathematically 367:10 368:17 276:3 281:5 360:10 363:1 125:1 142:22 mathematics 371:21 377:11 mechanicsm 383:3 384:13 144:21 153:7 mater 12:17 meaning 138:3 89:22 178:16 meeting 4:4,6 193:6 212:20 98:12 149:2 <td></td> <td></td> <td></td> <td></td> <td>· · · · · · · · · · · · · · · · · · ·</td>					· · · · · · · · · · · · · · · · · · ·
massive 195:2,5 232:9 244:19 measure 119:8 116:18 165:14 278:5 286:10 289:13 320:6 248:14 273:4 123:22 141:6 165:15 166:18 286:14 295:20 massively 277:3 282:1 221:11,18 167:18 168:16 347:20 352:13 335:20 289:10 290:20 345:18 169:8 171:9 378:21 383:6 match 85:9 91:8 291:3 321:11 measured 194:6,14,18 383:21 384:3 matching 12:19 327:18 332:15 291:18,18 196:7 203:4 mention 147:1 material 324:15 336:20 340:9 measures 23:21 264:2 268:9 210:6 220:14 math 235:11 340:10 353:6 23:22 68:5 284:17,22 290:3 270:1 283:13 354:13,18 112:8 163:22 354:6 355:7 mentioned 70:9 283:15 284:6 355:20 356:1 213:22 225:19 360:10 363:1 125:1 142:22 mathematics 371:21 377:11 377:22 mechanicism 383:3 384:13 144:21 153:7 matter 12:17 meaning 138:3 89:22 178:16 meeting 4:4,6 193:6 212:20				· · · · · · · · · · · · · · · · · · ·	
289:13 320:6 248:14 273:4 123:22 141:6 165:15 166:18 286:14 295:20 massively 277:3 282:1 221:11,18 167:18 168:16 347:20 352:13 335:20 289:10 290:20 345:18 169:8 171:9 378:21 383:6 match 85:9 91:8 291:3 321:11 measured 194:6,14,18 383:21 384:3 matching 12:19 327:18 332:15 291:18,18 196:7 203:4 mention 147:1 material 324:15 336:20 340:9 measures 23:21 264:2 268:9 210:6 220:14 math 235:11 340:10 353:6 23:22 68:5 284:17,22 290:3 270:1 283:13 354:13,18 112:8 163:22 354:6 355:7 mentioned 70:9 283:15 284:6 355:20 356:1 213:22 225:19 356:11 357:22 99:22 115:21 mathematically 367:10 368:17 276:3 281:5 360:10 363:1 125:1 142:22 mathematics 371:21 377:11 31:1 mechanicism 383:3 384:13 144:21 153:7 269:14 377:22 meaning 138:3 89:22 178:16 5:17 116:12 219:16,22					
massively 277:3 282:1 221:11,18 167:18 168:16 347:20 352:13 335:20 289:10 290:20 345:18 169:8 171:9 378:21 383:6 match 85:9 91:8 291:3 321:11 measured 194:6,14,18 383:21 384:3 matching 12:19 327:18 332:15 291:18,18 196:7 203:4 mention 147:1 material 324:15 336:20 340:9 measures 23:21 264:2 268:9 210:6 220:14 math 235:11 340:10 353:6 23:22 68:5 284:17,22 290:3 270:1 283:13 354:13,18 112:8 163:22 354:6 355:7 mentioned 70:9 283:15 284:6 355:20 356:1 213:22 225:19 356:11 357:22 99:22 115:21 mathematically 367:10 368:17 276:3 281:5 360:10 363:1 125:1 142:22 mathematics 371:21 377:11 31:1 meet 39:4 238:1 172:19 177:8 269:14 377:22 mechanism 33:4 31:11 meeting 4:4,6 193:6 212:20 80:4,5,8 90:21 248:15 277:21 212:10 263:15 5:17 116:12 219:16,22 98:12 149:2<	II ' ' I				
335:20 289:10 290:20 345:18 169:8 171:9 378:21 383:6 match 85:9 91:8 291:3 321:11 measured 194:6,14,18 383:21 384:3 matching 12:19 327:18 332:15 291:18,18 196:7 203:4 mention 147:1 material 324:15 336:20 340:9 measures 23:21 264:2 268:9 210:6 220:14 270:1 283:13 354:13,18 112:8 163:22 284:17,22 290:3 283:15 284:6 355:20 356:1 213:22 225:19 356:11 357:22 99:22 115:21 mathematically 367:10 368:17 276:3 281:5 360:10 363:1 125:1 142:22 mathematics 371:21 377:11 31:1 meet 39:4 238:1 172:19 177:8 269:14 377:22 mechanism 33:4 314:11 182:5 190:17 matter 12:17 meaning 138:3 89:22 178:16 5:17 116:12 219:16,22 98:12 149:2 326:3 263:22 319:18 117:1 193:15 221:16 223:2 98:12 149:2 326:3 322:16 346:17 193:18 203:6 239:21 243:14 253:22 294:18 145:19 190:19					
match 85:9 91:8 matching 12:19 291:3 321:11 measured 194:6,14,18 196:7 203:4 mention 147:1 383:21 384:3 mention 147:1 matching 12:19 material 324:15 matching 12:11 336:20 340:9 measures 23:21 264:2 268:9 270:1 283:13 340:10 353:6 23:22 68:5 284:17,22 290:3 264:2 268:9 290:3 290:3 290:3 210:6 220:14 290:14 290:3 270:1 283:13 354:13,18 283:15 284:6 355:20 356:1 213:22 225:19 356:11 357:22 354:6 355:7 mentioned 70:9 356:11 357:22 356:11 357:22 356:11 357:22 356:11 357:22 356:11 357:22 356:11 357:22 356:11 357:22 356:11 357:22 356:11 357:22 356:11 357:22 356:11 357:22 369:14 377:21 377:11 31:1 mechanicism 33:4 38:3 384:13 31:1 mechanicism 33:4 31:1 182:5 190:17 369:14 377:22 326:3 212:10 263:15 5:17 116:12 219:16,22 208:7 242:1 326:3 26:3 26:32 319:18 117:1 193:15 221:16 223:2, 208:7 242:1 meaningful 322:16 346:17 193:18 203:6 239:21 243:14 253:22 294:18 145:19 190:19 mechanisms 31:2 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20 matters 5:2 290:10 299:9 56:8 57:7,13 mellon 251:8 223:20 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8			,		
matching 12:19 327:18 332:15 291:18,18 196:7 203:4 mention 147:1 material 324:15 336:20 340:9 measures 23:21 264:2 268:9 210:6 220:14 math 235:11 340:10 353:6 23:22 68:5 284:17,22 290:3 270:1 283:13 354:13,18 112:8 163:22 354:6 355:7 mentioned 70:9 283:15 284:6 355:20 356:1 213:22 225:19 356:11 357:22 99:22 115:21 mathematically 367:10 368:17 276:3 281:5 360:10 363:1 125:1 142:22 mathematics 371:21 377:11 31:1 mechanicism 383:3 384:13 144:21 153:7 269:14 377:22 mechanism 33:4 89:22 178:16 meeting 4:4,6 193:6 212:20 80:4,5,8 90:21 248:15 277:21 212:10 263:15 5:17 116:12 219:16,22 98:12 149:2 326:3 263:22 319:18 117:1 193:15 221:16 223:2 98:12 149:2 326:3 232:16 346:17 193:18 203:6 239:21 243:1 253:22 294:18 145:19 190:19 322:16 346:17 193:18 203:6 239:21 243:1					
material 324:15 336:20 340:9 measures 23:21 264:2 268:9 210:6 220:14 math 235:11 340:10 353:6 23:22 68:5 284:17,22 290:3 270:1 283:13 354:13,18 112:8 163:22 354:6 355:7 mentioned 70:9 283:15 284:6 355:20 356:1 213:22 225:19 356:11 357:22 99:22 115:21 mathematically 367:10 368:17 276:3 281:5 360:10 363:1 125:1 142:22 mathematics 371:21 377:11 31:1 mechanicism 383:3 384:13 144:21 153:7 269:14 377:22 mechanism 33:4 314:11 182:5 190:17 matter 12:17 meaning 138:3 89:22 178:16 212:10 263:15 5:17 116:12 219:16,22 98:12 149:2 326:3 263:22 319:18 117:1 193:15 221:16 223:2 208:7 242:1 meaningful 322:16 346:17 193:18 203:6 239:21 243:1 253:22 294:18 145:19 190:19 mechanisms 295:22 347:4 246:1 249:22 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 <				, ,	
math 235:11 340:10 353:6 23:22 68:5 284:17,22 290:3 270:1 283:13 354:13,18 112:8 163:22 354:6 355:7 mentioned 70:9 283:15 284:6 355:20 356:1 213:22 225:19 356:11 357:22 99:22 115:21 mathematically 367:10 368:17 276:3 281:5 360:10 363:1 125:1 142:22 113:9 369:19 371:3,8 mechanicism 383:3 384:13 144:21 153:7 mathematics 371:21 377:11 31:1 meet 39:4 238:1 172:19 177:8 269:14 377:22 mechanism 33:4 meeting 4:4,6 193:6 212:20 80:4,5,8 90:21 248:15 277:21 212:10 263:15 5:17 116:12 219:16,22 98:12 149:2 326:3 263:22 319:18 117:1 193:15 221:16 223:2 208:7 242:1 meaningful 322:16 346:17 193:18 203:6 239:21 243:14 253:22 294:18 145:19 190:19 mechanisms 295:22 347:4 246:1 249:22 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 matters 5:2 290:10 299:	C		,		
270:1 283:13 354:13,18 112:8 163:22 354:6 355:7 mentioned 70:9 283:15 284:6 355:20 356:1 213:22 225:19 356:11 357:22 99:22 115:21 mathematically 367:10 368:17 276:3 281:5 360:10 363:1 125:1 142:22 113:9 369:19 371:3,8 mechanicism 383:3 384:13 144:21 153:7 mathematics 371:21 377:11 31:1 meet 39:4 238:1 172:19 177:8 269:14 377:22 mechanism 33:4 314:11 182:5 190:17 matter 12:17 meaning 138:3 89:22 178:16 5:17 116:12 219:16,22 98:12 149:2 326:3 263:22 319:18 117:1 193:15 221:16 223:2 208:7 242:1 meaningful 322:16 346:17 193:18 203:6 239:21 243:14 253:22 294:18 145:19 190:19 mechanisms 295:22 347:4 246:1 249:22 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 matters 5:2 290:10 299:9 56:8 57:7,13 meg 139:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20					
283:15 284:6 mathematically 355:20 356:1 367:10 368:17 213:22 225:19 276:3 281:5 360:10 363:1 125:1 142:22 360:10 363:1 37:21 377:38 369:19 371:3,8 371:21 377:11 377:22 377:21 377:11	math 235:11	340:10 353:6		284:17,22	
mathematically 367:10 368:17 276:3 281:5 360:10 363:1 125:1 142:22 113:9 369:19 371:3,8 mechanicism 383:3 384:13 144:21 153:7 mathematics 371:21 377:11 31:1 meet 39:4 238:1 172:19 177:8 269:14 377:22 mechanism 33:4 314:11 182:5 190:17 matter 12:17 meaning 138:3 89:22 178:16 meeting 4:4,6 193:6 212:20 98:12 149:2 326:3 263:22 319:18 117:1 193:15 221:16,22 208:7 242:1 meaningful 322:16 346:17 193:18 203:6 239:21 243:14 253:22 294:18 145:19 190:19 mechanisms 295:22 347:4 246:1 249:22 315:9 338:4 232:16 247:4 29:19 30:12 meetings 224:11 250:1 280:21 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 matters 5:2 290:10 299:9 56:8 57:7,13 meg 139:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20	270:1 283:13	354:13,18	112:8 163:22	354:6 355:7	mentioned 70:9
113:9 369:19 371:3,8 mechanicism 383:3 384:13 144:21 153:7 mathematics 371:21 377:11 31:1 meet 39:4 238:1 172:19 177:8 269:14 377:22 mechanism 33:4 314:11 182:5 190:17 matter 12:17 meaning 138:3 89:22 178:16 meeting 4:4,6 193:6 212:20 98:12 149:2 326:3 263:22 319:18 117:1 193:15 221:16 223:2 208:7 242:1 meaningful 322:16 346:17 193:18 203:6 239:21 243:14 253:22 294:18 145:19 190:19 mechanisms 295:22 347:4 246:1 249:22 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 matters 5:2 290:10 299:9 56:8 57:7,13 meg 139:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20	283:15 284:6	355:20 356:1	213:22 225:19	356:11 357:22	99:22 115:21
mathematics 371:21 377:11 31:1 meet 39:4 238:1 172:19 177:8 269:14 377:22 mechanism 33:4 314:11 182:5 190:17 matter 12:17 meaning 138:3 89:22 178:16 meeting 4:4,6 193:6 212:20 80:4,5,8 90:21 248:15 277:21 212:10 263:15 5:17 116:12 219:16,22 98:12 149:2 326:3 263:22 319:18 117:1 193:15 221:16 223:2 208:7 242:1 meaningful 322:16 346:17 193:18 203:6 239:21 243:14 253:22 294:18 145:19 190:19 mechanisms 295:22 347:4 246:1 249:22 315:9 338:4 232:16 247:4 29:19 30:12 meetings 224:11 250:1 280:21 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 matters 5:2 290:10 299:9 56:8 57:7,13 meg 139:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20	mathematically	367:10 368:17	276:3 281:5	360:10 363:1	125:1 142:22
269:14 377:22 mechanism 33:4 314:11 182:5 190:17 matter 12:17 meaning 138:3 89:22 178:16 meeting 4:4,6 193:6 212:20 98:12 149:2 326:3 263:22 319:18 117:1 193:15 221:16 223:2 208:7 242:1 meaningful 322:16 346:17 193:18 203:6 239:21 243:14 253:22 294:18 145:19 190:19 mechanisms 295:22 347:4 246:1 249:22 315:9 338:4 232:16 247:4 29:19 30:12 meetings 224:11 250:1 280:21 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 matters 5:2 290:10 299:9 56:8 57:7,13 meg 139:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20	113:9	369:19 371:3,8	mechanicism	383:3 384:13	144:21 153:7
matter 12:17 meaning 138:3 89:22 178:16 meeting 4:4,6 193:6 212:20 80:4,5,8 90:21 248:15 277:21 212:10 263:15 5:17 116:12 219:16,22 98:12 149:2 326:3 263:22 319:18 117:1 193:15 221:16 223:2 208:7 242:1 meaningful 322:16 346:17 193:18 203:6 239:21 243:14 253:22 294:18 145:19 190:19 mechanisms 295:22 347:4 246:1 249:22 315:9 338:4 232:16 247:4 29:19 30:12 meetings 224:11 250:1 280:21 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 matters 5:2 290:10 299:9 56:8 57:7,13 meg 139:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20	mathematics	371:21 377:11	31:1	meet 39:4 238:1	172:19 177:8
80:4,5,8 90:21 248:15 277:21 212:10 263:15 5:17 116:12 219:16,22 98:12 149:2 326:3 263:22 319:18 117:1 193:15 221:16 223:2 208:7 242:1 meaningful 322:16 346:17 193:18 203:6 239:21 243:14 253:22 294:18 145:19 190:19 mechanisms 295:22 347:4 246:1 249:22 315:9 338:4 232:16 247:4 29:19 30:12 meetings 224:11 250:1 280:21 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 matters 5:2 290:10 299:9 56:8 57:7,13 meg 139:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20	269:14	377:22	mechanism 33:4	314:11	182:5 190:17
98:12 149:2 326:3 263:22 319:18 117:1 193:15 221:16 223:2 208:7 242:1 meaningful 322:16 346:17 193:18 203:6 239:21 243:14 253:22 294:18 145:19 190:19 mechanisms 295:22 347:4 246:1 249:22 315:9 338:4 232:16 247:4 29:19 30:12 meetings 224:11 250:1 280:21 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 matters 5:2 290:10 299:9 56:8 57:7,13 meg 139:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20	matter 12:17	meaning 138:3	89:22 178:16	meeting 4:4,6	193:6 212:20
208:7 242:1 meaningful 322:16 346:17 193:18 203:6 239:21 243:14 253:22 294:18 145:19 190:19 mechanisms 295:22 347:4 246:1 249:22 315:9 338:4 232:16 247:4 29:19 30:12 meetings 224:11 250:1 280:21 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 matters 5:2 290:10 299:9 56:8 57:7,13 meg 139:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20	80:4,5,8 90:21	248:15 277:21	212:10 263:15	5:17 116:12	219:16,22
253:22 294:18 145:19 190:19 mechanisms 295:22 347:4 246:1 249:22 315:9 338:4 232:16 247:4 29:19 30:12 meetings 224:11 250:1 280:21 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 matters 5:2 290:10 299:9 56:8 57:7,13 meg 139:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20	98:12 149:2	326:3	263:22 319:18	117:1 193:15	221:16 223:2,9
253:22 294:18 145:19 190:19 mechanisms 295:22 347:4 246:1 249:22 315:9 338:4 232:16 247:4 29:19 30:12 meetings 224:11 250:1 280:21 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 matters 5:2 290:10 299:9 56:8 57:7,13 meg 139:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20	208:7 242:1	meaningful	322:16 346:17	193:18 203:6	239:21 243:14
315:9 338:4 232:16 247:4 29:19 30:12 meetings 224:11 250:1 280:21 358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 matters 5:2 290:10 299:9 56:8 57:7,13 meg 139:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20	253:22 294:18	\circ	mechanisms	295:22 347:4	246:1 249:22
358:11 379:11 263:18 282:2 31:6 55:10 253:20 293:13 319:10 matters 5:2 290:10 299:9 56:8 57:7,13 meg 139:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20		232:16 247:4	29:19 30:12	meetings 224:11	250:1 280:21
matters 5:2 290:10 299:9 56:8 57:7,13 meg 139:10 mentioning 118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20		263:18 282:2	31:6 55:10	0	293:13 319:10
118:21 124:5 301:1 329:21 63:17 86:1 mellon 251:8 223:20					
			,	\cup	<u> </u>
∥ 215:14 338:5 331:11 93:1.8.20 95:2 member 6:10.11 mentions 89:1	215:14 338:5	331:11	93:1,8,20 95:2	member 6:10,11	mentions 89:1
mature 126:21 meaningfully 101:15 173:22 41:18 102:6 mere 78:14			, ,		
maximize 61:17 333:2 210:22 211:19 110:11 134:14 170:6		_ ,			
mean 9:13 11:7 meaningless 212:4,16 141:20 192:2 merely 232:21					
13:9 23:9,10 328:12 244:22 306:4 203:8 226:8 371:18		0	,		•
30:13 44:11 means 11:20 308:8 316:3,4 255:2 merge 34:2	II '				
46:21,22 52:7			,		0
	II '				34:19 36:21,22
69:12,17 78:6 79:3 124:1,2,4 342:5,15 6:13 7:9 9:18 merges 34:13				· · · · · · · · · · · · · · · · · · ·	,
78:16 84:18 187:16 189:4 media 338:14,22 9:21 10:12 merging 34:18	ll '	, ,	,		0
87:10 102:19			· · · · · · · · · · · · · · · · · · ·		0 0
	07.10 102.17	202.3 221.17	370.3	25.10 20.15,14	J T .21 JJ.2

37:3,16	346:3	181:17,21	267:4	modified 88:3
merit 362:2	midst 316:14	minister 166:12	misuses 266:22	modify 212:4
380:12	mike 124:11	minorities 122:2	mit 251:8	modifying 85:5
meritorious	134:8 144:13	minute 109:2	mitigate 46:15	moment 316:2,2
362:17 365:20	149:5 167:19	110:2 140:3,3	92:20 225:2	323:20 324:1
merits 113:5	185:21	159:16 201:9	232:17 261:8	325:10 347:11
mess 8:7	mikes 164:20	217:11 284:18	262:2	347:11 362:6
messages 136:8	military 122:21	353:18	mitigated 20:21	369:18 380:17
messy 117:17	123:2 227:16	minutes 9:10,12	66:9	momentum
metadata 37:10	228:4	9:16,17 41:16	mitigating 93:1	178:14
102:12 103:4	million 361:1	74:11 76:1	93:4 280:17	monday 224:18
152:1,4,11,12	mimic 353:3	109:21 139:7	mitigation 92:19	money 342:20
190:18,18	mind 56:5 82:20	141:19,20	mobile 135:21	monitor 122:16
metaphor 260:7	101:13 121:8	205:20 217:7	136:4	monitoring
method 106:12	145:13,16	278:10 286:1,5	mode 58:12	343:21
160:8 218:11	180:22 201:4	286:9,10	model 35:21,22	monitors 22:12
267:11	306:6 318:15	misbehavior	83:5 115:12	months 95:14
methodologies	336:7 346:9	27:21	118:14,19	311:20 379:11
237:5	366:12	misconception	121:9,11,13	moot 78:14
methodology	mindful 146:14	119:3	135:10 155:7	moral 121:18
59:4	182:4	misconduct	155:19 157:22	122:5
methods 35:5,10	minds 148:15	31:15 192:15	modeling	morality 27:18
40:18 62:16	mine 199:15	misinterpreted	111:14 114:15	morning 4:2
89:8 91:21	259:7 271:8	22:16	114:19 115:19	18:5,6 109:9
92:6 182:21	miniminization	mismatch	182:9 183:3	109:10 110:19
214:18 218:16	272:12	351:13	199:20	117:9 184:17
219:3 267:2	minimization	mismatched	models 114:19	202:22 203:13
metric 165:7	118:7 126:1	346:18	182:19,22	203:19 204:7
mic 318:16	221:16 233:8	misplaced 13:18	183:2 261:11	234:1 245:11
323:21	297:7,16 300:4	missed 173:5	261:12	255:21 268:12
michael 3:6	300:5,6 301:18	184:12	moderate 193:7	285:13 287:14
190:16	327:14 328:1	mission 216:20	moderated 6:9	mortal 313:20
micro 28:22	329:9 358:9,14	227:5,22	7:19 203:8	mosaic 36:13,18
microphone	359:3,4 372:4	298:13 304:13	moderating	47:11 75:13
203:15,18	376:2	383:18	109:6 285:8	76:7 77:9 78:5
375:15	minimize 106:5	missions 215:15	moderator 6:17	80:11 89:1,2
microsoft 3:6	204:20 361:13	298:7	9:16 286:8	89:10,12,18
124:12 125:2,3	382:10	mistake 55:19	modern 32:11	97:5 383:14
144:14 149:7	minimized	93:16 316:15	97:21 186:4	mosiac 268:13
149:10 150:17	358:3,5	mistakes 173:11	329:11	mother 13:8,12
150:17,20,22	minimizing	174:17 213:9	modest 58:19,21	motivated 99:12
151:1,2	62:14	misuse 56:21	91:4	motivations
mics 318:17	minimum 62:20	86:6 94:18	modifications	44:9
middle 92:18	mining 47:9	misused 55:17	85:14 374:12	motive 44:17
		1	I	1

l 				
mount 141:5	201:13 202:15	257:1,10 280:1	208:21 259:17	312:5,7,9,12
170:13,20	nail 150:12	285:7 294:10	290:11 300:18	312:3,7,9,12
move 19:10 96:6	naive 166:17	296:3 297:20	301:22 317:8	317:13 310.5,0
116:6 236:16	naivete 355:15	297:22 298:8	345:5,6 370:16	317:13 319:0
270:9 301:6	name 5:10	298:18 299:8	necessities 86:7	320:14 321:17
383:18 384:10	237:13 306:20	302:5,7 310:19	necessity 54:17	331:5 345:15
movements 15:2	names 175:13	302.3,7 310.19	204:11 208:9	347:7 348:2,3
148:8	194:9	331:3,12	317:6 345:3	348:5,5 349:5
		,		· ·
movie 160:17	nara 222:5	335:11 367:9	362:14 366:3	349:12,15
moving 89:21	narrow 18:15	375:19 381:21	necessarily	354:7 355:17
94:15 96:15	19:10 53:15	383:19	218:21	355:21 359:12
219:7 305:16	65:4 72:3	nationality	need 23:11	360:8,9 373:2
318:15 369:10	365:12	231:9	24:20 26:6	377:18
370:13	narrower 360:1	nations 111:22	28:22 29:6	needed 40:21
multifaceted	narrowing	112:19 325:20	32:1 46:2 47:1	118:14 217:14
173:14	300:14	365:18	47:21 52:13,15	243:18 282:21
multihop 90:22	narrowly 31:5	natural 26:18	53:14,16 62:10	300:15,20
multilaterally	45:18 366:1	106:15	63:4 66:7 67:1	307:4 359:7
153:13	nation 23:4	naturally	67:4 68:16	needing 95:22
multilayered	111:2 115:2	378:21	83:10 89:11	needle 257:5,7
173:15 358:20	116:13 189:19	nature 32:15	90:3,4 93:17	needs 39:4
multinational	209:3 226:11	58:11,22	93:20 94:3,10	54:16 66:2,3
343:13	226:21 227:11	188:19 226:14	105:4 113:22	67:15 85:9
multiple 308:20	228:4,6,9	229:1 230:19	119:18 127:21	90:17 116:13
308:20 360:19	234:20 302:2	333:18 336:1	130:10 132:8	128:2 133:8
370:5	national 3:11,14	379:6	133:6 134:3	134:1 167:17
multivaried	8:17 10:8 46:2	naval 313:10	143:15 146:13	179:18 193:1
32:15	53:22 64:6	navigate 335:17	146:19 155:4,7	195:9 204:5
music 135:8	66:13,18,20	near 308:5	168:12 170:9	236:17 254:2
muslim 16:18	67:4,5,6,9	nearly 124:21	170:17 172:7	259:1,10
muster 332:4	69:10 81:1,12	necessarily	174:18 175:2	273:20 281:6
mutually 155:21	84:13 85:9	24:15 98:12	176:12 183:1	311:3 374:8
296:22	89:9,13 104:10	155:4 168:21	187:11 190:9	nefarious 112:2
mysterious	111:15 112:18	182:8 184:8	193:16 194:4	negative 20:10
369:9	123:22 129:14	190:9,15	198:8 200:13	35:19 175:7
mythical 331:7	129:18 133:20	218:20 275:15	210:16 214:8	204:20 291:8
	162:16 176:13	279:2 284:7	214:21 234:17	291:17 312:22
N	181:15 200:11	305:9 315:4	245:5 254:18	381:17 382:8
n 1:13	204:5 206:6,12	333:13 338:11	257:4 261:21	negatives 56:12
nahari 3:7	218:15 219:19	365:4 376:8	265:2 267:4	negligent 95:10
134:15,20	222:6 223:2,3	necessary 31:7	270:19,19	neighborhood
154:17 155:1	223:4 225:11	40:20 55:3	283:15,16	183:12 184:20
158:11 169:22	225:22 226:15	57:22 161:13	294:11 309:21	neighbors
182:3 189:9	246:19 256:21	178:11 192:8	310:3,8,9	148:22
	<u> </u>	<u> </u>	<u> </u>	

neither 44:17	84:19	89:21 99:1	nsas 15:14 23:14	obligation 86:20
nervous 80:8	noncontent	155:16 156:10	225:17,20	240:13
355:8	152:1	167:11 190:17	226:6 227:2,5	obligations 38:9
netflix 160:15	nonexistent	201:17 289:15	227:18 229:14	obscure 14:20
161:10 195:11	307:6	363:6 370:20	229:17,20	15:5
network 197:16	nonintelligence	374:1	231:1,15	obscurity 5:3
199:7 369:16	231:17	notions 196:20	314:20 351:18	15:8 74:19
369:16	nonpregnant	197:6	nuance 29:1	76:19,22 77:5
networks	35:21	notwithstandi	348:21	observation
131:10 197:22	nonstate 228:7	369:8	nuanced 307:20	15:4 27:14
neutral 374:2,16	nonstatutory	novelist 8:3	368:5	30:1 78:9
380:4	300:9	november 1:7	nuclear 60:4	observations
never 55:16,16	nontechnologist	4:5	227:12	176:10 257:19
55:18,19	137:7	novo 53:6	nudge 39:2	287:22
120:16 164:1	nonu 132:9	nowadays 40:7	number 17:13	observe 293:20
303:17 326:5	282:9 301:8	nsa 3:20 13:11	37:5 66:4	observed 5:5
351:7 354:1	302:11	81:4 82:5	141:6 149:19	79:10 337:15
nevertheless	noon 7:5	84:11 103:8	153:21 164:5,7	obtain 81:20
131:11	normal 15:4	104:7 130:20	164:9 172:8,9	131:2 221:8
new 25:15 47:8	78:9	164:3 165:3	172:11 173:6	227:5
50:18,18 76:14	normative 167:4	172:2,4 177:3	189:19 198:13	obtained 223:11
78:18 118:14	notable 234:8	177:9 225:15	200:17 287:4	266:19 280:11
118:19 132:16	notarial 385:12	227:14,22	320:19 363:4	352:4
132:17 140:9	notary 385:4,17	228:1,8,10	numbers 91:12	obtaining 261:4
143:4 145:18	note 6:21 28:14	230:3,10,14	91:19 175:15	262:19 265:8
157:14 158:15	75:22 78:3	231:20 232:4	175:15 194:9	266:6
172:2,3 177:8	116:1 179:13	233:3,6 234:2	229:10	obvious 11:19
190:9 191:9,11	noted 81:2	238:10,13	numerous 88:22	39:22 42:14
209:20 225:13	358:1	248:17 250:14	nuts 96:1 354:12	348:11
230:4 231:4	notes 61:9	251:1,17,22,22	nvidia 3:7	obviously 28:2
241:9,9,10	nothings 173:13	269:13 270:16	134:16	43:5 63:16
250:1 260:13	notice 83:8 84:2	275:12 276:8	nypds 16:18	81:7 89:2 95:3
260:13 271:22	84:17 272:5	283:13 284:1,6		99:17,18
273:2 330:18	289:11 290:6	301:13,15	0	102:20 150:13
364:9	290:10 312:14	313:6 314:14	oath 244:4	154:15 163:7
newer 340:5	328:6 329:5	316:13 319:14	obama 132:7	176:9 194:14
newly 209:12,18	332:13,19,21	320:8 323:1	obfuscate 68:7	194:21 204:2
news 16:17	343:4	325:16 330:11	objected 339:5	221:6 223:18
163:19	noticed 203:13	333:21 351:6	objection	241:15 272:7
nice 225:9	326:12	351:15 357:4	339:14	272:16 273:9
235:12 325:8	notified 84:6	360:16,21	objective 86:5	276:20 297:5
nodding 97:10	notion 8:2 55:4	364:2,13,19	219:2 223:12	311:8 341:22
98:2 107:18	69:7 80:17,20	365:9,14	objectives 27:5	342:11 350:11
nofly 83:14	81:11 82:3	380:16	obligated 309:9	367:11,12
	I	1	I	I

occasion 148:5	223:8 244:5,8	364:8	228:2,11,14	opposition
occur 37:3	244:11 264:5	older 322:10	operates 130:4	69:18
occurred 228:21	358:22	omg 354:8	operating	option 197:18
occurring	official 224:10	357:20 358:19	210:21	328:10
185:18 187:22	239:22	359:21	operation	options 39:22
240:16	officials 3:10 6:4	once 29:16	204:12 374:14	90:8,12 308:3
occurs 110:8	8:21 15:13	157:19 186:3	operational	308:4,6
184:1	23:7 66:5	191:1 258:22	28:20 32:2	oral 120:3
october 4:10	204:1 205:5	293:8 300:8	85:9 86:7	318:21 326:18
odd 294:11	215:21 232:10	321:1 351:1,2	218:19 315:18	order 4:16
odds 69:7	383:9	ones 5:2,6,10	operations	68:10 70:12
315:12	oftentimes 48:1	13:4 42:2,10	215:15 217:20	90:5,18 109:18
odni 239:1	218:21 358:5	51:11 57:17	217:22 225:9	130:4 159:17
244:20,21	oh 136:9,11	83:19 85:8	227:16 313:11	196:13 208:12
249:15	187:8 189:11	94:20 163:7	355:17 373:9	217:13 221:8
odnis 206:10	199:6 201:11	210:3 252:17	operators 239:6	221:14 229:18
oecd 329:4	207:3 262:6	272:10 273:17	316:18 334:15	238:14 254:2
offensive 111:1	305:2 355:3	279:16	334:16	254:18 257:5
offer 32:14	ohm 195:12	ongoing 142:19	opine 253:6	259:1 265:2
33:18 38:14	ohms 311:11	263:16,16,22	opinion 80:3	279:1 301:5
63:11 86:22	okay 18:1 33:8	online 16:10	148:6 199:12	323:9 324:14
257:19 287:21	41:15 42:13	34:9,10 39:12	368:19 369:9	orders 128:9,9
offers 39:21	45:10 52:4	39:19 69:1	375:2 380:5	128:20 129:18
291:13 357:15	67:17,21 82:21	125:8 151:10	opinions 301:22	129:22 169:3
office 3:11 206:6	83:3 87:7	152:9,13	374:10	301:22 325:18
206:11 215:16	88:11,13 102:1	160:16 161:10	opportunities	ordinary 128:13
215:22 238:3,4	102:2 103:21	onset 177:17	91:8 226:19	228:17
238:5,7,10	105:9 106:14	ontological	opportunity	organic 326:3
240:1,6 243:14	106:17 108:16	26:15	6:12 25:11	organization
243:15 244:6,7	134:11 159:14	onus 63:5	33:17 96:5	115:13 234:3
245:4 246:5,19	161:8,9 163:1	open 54:16 55:3	110:20 117:10	244:21 252:4
351:18 354:20	164:14 167:18	107:14,16	124:16 134:20	264:21 265:7
officer 4:11	179:22 180:16	219:10 286:11	230:22 237:14	268:3 308:12
120:12 171:18	194:14 201:4,8	318:6	237:18	320:8 357:17
172:3,4 206:5	207:7 255:8	opening 4:17	opposed 43:22	357:19
206:21 215:11	257:18 259:6	109:15,20	44:19 54:8	organizations
223:18 225:13	267:13 274:7	157:22 172:6	70:22 192:19	34:7 38:7
226:7 239:12	275:20 277:7,8	237:21 340:10	251:6 263:19	106:19 107:6
officers 56:20	305:2 354:14	openly 122:20	277:9 287:22	232:2 264:19
203:2 243:7	357:15 369:14	123:12	318:5 338:8	304:13 310:8
264:18 265:10	369:20 383:1	operate 95:22	355:22	320:16
349:12,14	old 106:21	126:10,18	opposite 11:21	organizing 63:8
offices 209:11	117:20 137:22	208:9 346:8	214:14 243:10	orient 217:1
209:11,13	247:12 259:7	operated 204:19	355:20	orientation
	l	l	1	Į.

				1
216:15	oversaw 303:13	338:1 342:21	171:13 179:9	paradoxically
oriented 114:18	overseas 229:22	349:12,13	181:10 185:10	361:15
originally	oversee 41:4	375:18 376:2	199:10 200:5	paralyze 379:3
175:13	254:18	383:13	202:22 203:1	pardon 119:19
ought 42:11	oversees 215:17	oversights 30:10	203:21 204:22	parentage
63:15 75:18	oversight 1:3	56:8	255:21 256:1	303:20
229:7 344:17	4:3 21:19	overtime 327:12	269:11 276:20	parents 303:21
359:6	24:12,17 30:11	overusing	278:5,10,17	parker 7:15
outcome 385:11	31:19 38:4,22	190:17	285:3,4,13,16	286:15
outcomes	41:7 42:7	overweighing	285:17,17,18	parking 137:1
291:10	52:12,16 71:12	73:13	287:14 288:3	part 11:20 49:9
outcry 298:15	72:19 73:2,18	overwhelming	288:18 310:5	55:11 65:16
outdated 25:21	73:19 94:11	144:17 319:4	319:11 333:12	67:13 82:10
outlined 101:21	95:21 96:5	ownership 70:3	333:13 341:17	83:22 86:3
outputs 143:14	130:9 157:2		347:20 363:6	87:10 107:22
outset 332:14	165:10 169:4	P	363:22 383:4	108:17 120:10
outside 10:21	174:17 177:5	pace 341:8	panelist 9:9	137:11 155:9
13:17 29:18	178:1 179:15	page 241:10	165:16 252:22	172:1 178:15
31:4 43:5	192:4,6,14,18	pages 119:13	286:1 295:14	207:15 210:14
57:20 59:4	193:2 194:5	163:1 220:8	303:9 313:4	212:8,15,19
94:16 95:4	203:5 207:13	paint 36:15	panelists 6:17	214:1 215:16
116:19 194:1	209:6,7,17,19	pairing 173:22	6:18 7:1 9:3	216:20 224:22
209:13 230:14	210:1,6,7,10	paladin 3:20	10:4 25:19	225:4 227:5
353:7 355:13	210:12,14	313:5	64:17 79:7	232:19 233:6
369:18 373:3	211:1,18	palantir 303:10	88:7 105:11	240:15 241:2
373:20,21,21	212:16 219:15	304:2,4,11,20	154:13 185:10	243:18 244:4
overall 112:19	219:16,20	305:19 309:9	203:7 206:3	247:8 250:5,12
114:5 200:8	223:2 225:17	309:11 311:10	275:21 285:10	262:20 274:20
261:13 262:3	238:22 239:19	320:15 338:19	286:2 313:12	277:3 280:22
263:2 301:6	241:17 244:8	349:4 367:8	318:10,14	289:8 291:5
overarching	246:1,3,17,18	palm 136:11	384:3	298:12 320:13
23:3 49:21	247:4 252:18	panel 6:6,9,19	panels 5:20	339:2 346:4
overcome 68:20	253:2,10,14,15	7:18,20 10:2	116:5 140:7	372:5,7
overlapping	253:21 254:3,6	26:14 31:14	178:12 179:9	partial 71:6
296:22	254:22 276:16	32:5 41:17	199:10 206:16	participate
overlook 331:13	276:19,21	69:22 85:21	286:13 313:18	10:13 124:17
375:7	277:17,21	88:16,18 96:10	383:4	299:6
overlooked	295:21 306:4	102:6 108:17	paper 271:13	participated
173:5	307:5,15 308:8	108:22 109:3,7	276:18 306:13	360:17
overly 353:11	318:19 319:5,7	109:10,11	319:1	participates
overrepresent	319:9,12,15,18	110:8,12 119:3	papers 12:21,22	362:18
184:2	319:22 321:12	134:14 135:6	42:3	participation
override 65:2	321:14,19	142:22 155:3	paperwork	19:3 80:22
overriding 12:6	322:16 324:21	164:21 171:11	192:12	82:3,9,16 86:3
	•	-	•	•

163:8 221:5	94:13,15	10:2,17 11:22	252:19 255:20	70:10
271:18 272:8	230:17 233:12	15:20,21 16:4	260:4 262:19	perception
274:9 326:15	250:17 273:15	16:7 17:3,7,8,9	263:5,17	253:4
326:22 327:10	360:13 371:19	17:13,19 18:9	264:18 266:18	perceptively 8:4
327:19 328:14	372:19	,	267:2,19	
		18:10 19:1,2	,	perfect 111:11
329:9,18	party 150:16	20:9 21:5,22	278:19 283:13	114:4,7,12
particular 9:8	151:6 168:17	22:6,17 23:5	284:5 294:18	115:18 133:18
21:15 29:2	361:22,22	23:17,20,21	294:20 303:17	160:2,5,10
37:8 40:19,22	pass 187:20	24:3,6,7,7,9,16	304:2 309:21	161:21 162:3
41:18 49:2	286:19 332:4	34:4,7 35:6	310:3 312:9,11	173:13 194:21
65:10 66:21	364:7	37:1,21 43:2	319:13 321:15	213:8 241:12
67:2,2 86:17	passed 145:8	44:10,13,18	321:15,17	perfectly 70:4
94:1 102:6	230:16	47:19 48:11,17	322:1,4 328:18	100:10 354:11
121:22 129:5	pat 255:7	49:12 54:17	329:14 330:15	perform 38:10
181:1 192:1	path 276:11	56:1 58:1,16	332:18 338:16	104:22 254:3
204:13 208:4	paths 91:9	58:20 59:18	339:10,20	254:22
211:15 214:4,9	patricia 2:5 4:15	64:2 65:22	342:22 345:11	performance
216:13 220:14	patriot 103:5	68:1,5,8 69:20	346:14 352:4,8	134:17
225:10 240:12	301:12	70:1 71:5,6,9	354:22 355:1	performed
241:3 245:2,19	patternbased	74:8,18 76:15	357:20 361:1	233:11
248:17 249:14	181:16,20	79:3 81:18	363:17 368:3	performing
280:9 281:16	paul 2:14 25:4	83:6 95:8 98:6	379:17 381:2	254:17
286:16 315:14	64:12 68:15	99:3 104:2	peoples 12:4	period 138:17
323:19 328:1	69:4 98:2	113:14 120:12	18:19 21:3,4	196:11 206:2
333:20 335:9	100:18 106:14	120:18 122:2	23:16,17 42:20	periodically
particularily	195:12 311:11	123:8 132:1	45:5 47:18	241:8
278:9	pauls 40:4	133:9 146:9	53:9 57:2,5	periods 15:2
particularly	pausing 81:15	147:21,22	59:10 128:13	93:2
47:5 109:10	pay 243:8 286:6	148:5,17,21	148:15 150:9	permanently
124:7 151:13	287:12 336:16	149:16,18,20	151:16,18	196:16
168:2 203:14	paying 157:4,5	151:8 152:20	153:3 198:17	permissable
210:8 240:22	pays 190:7	153:8 160:18	209:2 373:10	238:17
244:6 245:8	pbgc 106:20	163:18 166:22	379:12	permission
246:3 285:19	pclob 110:6	172:16,17,20	perceived 213:6	348:10
289:16 294:10	114:9 116:2	173:7 178:5	315:13 317:19	permit 29:10
334:16 342:7	159:9 279:18	180:7 183:15	percent 16:9	299:9
parties 42:17	295:6 303:1,19	184:13 197:3,4	39:18 43:3	permitted 130:2
63:13 91:3	318:8 353:6	197:15,18	56:4 70:12	222:16 243:2
151:18 153:22	355:6,10	198:14,20	93:17 157:10	294:3 314:22
375:9 376:21	358:22 361:6	199:5,22	157:10 159:13	permitting 6:16
385:10	382:21 384:9	200:19 214:7	163:19 196:22	perpetrators
parting 108:18	penalties 222:22	242:11,20	198:16 363:13	183:16
parting 100.16 partner 313:5	pension 107:2,3	243:11 247:1	379:1	persecute 44:10
parts 47:11	people 6:15 9:22	249:11 251:7	percentage	persecution
Period 17.11	People 0.13 7.22	2 17.11 2J1.1	Percentuge	Persecution

55:18	33:19 94:15	phones 112:3	335:16,20	70:7 78:13,14
person 12:20,21	97:7,8 124:19	135:22 139:5	345:11 352:5	79:4,8 81:22
15:9 34:15	135:7 137:12	363:17	357:10.10	91:10,10 93:6
79:12 97:14	184:6 208:8	phrase 363:4	361:8 364:19	96:19 100:22
98:9 103:6	216:17 219:6	366:9		101:5 102:12
			placed 39:17	
120:14 150:15	220:7 229:14	phrasing 260:21	376:20	108:2 120:4
230:2 238:19	244:14 251:5	physical 336:21	places 5:8 29:15	123:15,17
240:6 256:13	254:15 265:17	physically	108:8 149:1	142:2 144:14
256:15 269:5	266:16 270:12	170:10	305:4 311:12	146:22 147:15
278:22 279:3	270:13 273:14	physicians	planning 159:22	159:15 162:21
289:20 294:16	277:17 280:1	122:22	platform 220:17	179:12 183:2
311:1 376:8	280:15 334:10	pick 106:16	304:5	189:2,10
personal 5:1,11	338:14 360:16	297:3 326:20	platonic 49:22	192:13 238:11
14:10 17:10	360:21	picking 203:18	72:2	245:3 246:1
20:13 27:18	perspectives	picture 36:16	play 73:21 74:7	256:19 257:2,9
43:12,21 51:11	232:9	123:7 156:16	74:7,9 83:12	257:9,22 258:6
118:21 127:8	pertain 315:6	156:20 170:18	172:17 272:21	260:3 275:15
133:9 136:9	317:4	249:13 261:13	373:8 375:18	277:12 286:8
228:18 229:1	pertaining	piece 64:8	377:1	290:8,13 292:1
231:11 232:3	225:16	203:12 222:3	players 209:9	300:7 305:22
235:4,14,17	pervasive 30:22	256:13 324:2	playing 73:21	307:17,19,19
236:7,7,11,13	341:2	371:5	239:1	308:8 310:14
242:17 254:15	pervasiveness	pieces 47:5,15	plays 12:4	310:14 315:8
265:8 368:19	30:19 186:10	76:12 263:20	315:16	316:22 319:11
personalities	pessimistic 8:8	306:13	plcob 7:8	323:14 324:13
54:18	peter 6:14 7:16	pierce 15:7	pleasantly	335:9 350:7,9
personality	ph 110:14	pike 341:1	242:19	354:7 356:15
37:13	145:19 284:6	pilot 249:22	please 4:17	361:14 367:12
personally	phase 78:22	pioneers 205:8	110:18 134:19	371:8 377:6
85:18 160:21	218:7 258:1	pipeline 34:1,13	154:14 201:12	pointed 64:12
178:3 229:8		35:4	286:6	75:12 195:11
243:5 247:16	346:1,21	:		
	phases 37:17,17	place 42:18 96:2	plenty 16:4	252:18,22
249:10	255:19	137:1 144:20	78:21	pointing 306:5
personnel	phenomenon	178:5 180:4	pllc 2:14	points 18:6 71:4
126:22 127:6	142:10	209:22 232:3	pockets 139:6	76:18 77:13
persons 15:1	phishing 339:3	233:14 242:3	point 15:12 16:2	118:20 252:12
29:15 66:21	phone 37:10	245:19 252:8	16:3 18:7	277:21 307:13
67:2 78:10	90:15 91:3,12	267:3 269:5	19:12 20:12	309:4 346:1,21
113:21 118:22	91:18 135:21	270:22 272:7	22:21 23:19	police 56:20
131:18 132:1	136:1,4,5	273:11 275:3	35:15 39:21	57:3,5 120:12
164:5 231:7,10	166:8 170:12	275:14 288:22	49:21 51:21	183:10 185:2
279:3,21 282:9	194:8,12	300:2 323:7	57:19 60:9,14	policies 38:8
301:8 302:11	278:19 301:14	330:10,22	60:17,20 66:12	83:7 95:20
perspective	301:16	331:7 333:1	66:17 69:5	119:9 125:15

<u></u>				4.
105.16.120.2	160-6 206-21	07.14.110.17	242.6 272.16	27.10 102.12
125:16 129:3 144:20 145:8	169:6 206:21	87:14 118:17	242:6 373:16	37:12 183:13
	216:18 238:13	121:10	practice 38:1,16	predictive 35:5
214:15 215:4	239:18	posted 384:8	63:15 100:2	35:17,21,22
216:5 230:18	positions 242:12	postindustrial	106:6 114:10	182:9 183:3,13
234:14 338:21	264:11 268:1	76:19	146:7 204:9	184:9,15
348:5,12 367:1	positive 27:2	posting 191:16	205:8 287:19	preerror 95:2
policy 25:8	35:18 128:1	posture 111:22	296:9 303:3	preferences
33:14 41:3,7	129:4 229:6	potential 30:18	314:4 315:18	33:5
80:4,5,8	291:10,16	115:7 142:14	354:9 364:12	pregnancy 35:9
110:17 113:4	293:22 354:19	144:7 169:17	practices 33:20	pregnant 35:20
113:12 115:5	positives 56:12	174:15 234:18	33:22 39:7	37:9
142:13 145:2,3	56:14 58:16	265:4 268:20	62:8 63:6	prem 286:16
174:13 175:5	positoin 273:8	291:17 348:8	108:13 111:21	premise 96:21
240:12 251:4	possess 305:9	381:17 382:7	112:12 125:21	96:22 98:16
257:8,14	possessing	potentially	193:1 197:22	158:8 186:13
259:22 282:10	302:15	222:20 253:1	231:21 296:20	186:14 282:18
315:5 322:6,11	possession	370:17	299:11	283:1,3,7,9,12
327:6 329:17	202:18	power 21:12	practicing	313:18
339:3 342:3	possibilities	36:18 44:7	267:15	premised 98:22
349:11 350:4,5	169:19	122:3 302:14	practitioners	premises 81:16
352:21 360:15	possibility 56:17	302:16,21	316:18	prepared 7:7
362:6,8 380:6	56:19 87:5	306:7 337:5,12	pre 70:22	63:10
polishing	110:4 141:21	337:22 357:17	pre 2007 137:22	preprocessing
182:19	142:18 345:15	powerful 117:15	precedent	62:14
political 44:14	365:21	124:5 163:10	150:14	prescribed
45:3 122:3	possible 7:17	167:6 337:13	precise 34:19	362:21
politically 44:19	56:10 57:1	337:14 357:11	35:2 168:7	prescriptive
327:5	118:2,3,15	powers 15:4	307:11	221:18
poll 16:8 197:1	143:14 156:19	78:9 111:16	precisely 99:10	present 4:12
363:12	157:1 174:16	314:7 375:9	182:13 307:19	68:10 227:3
pollution 20:19	235:6,16 236:3	378:5	precision 34:21	268:16 336:22
pool 46:10	236:6,8 348:18	ppd28 231:5	38:15	presentation
pop 269:19	349:16 378:8	practicable	predicated	380:18
population 36:4	possibly 102:7	162:7	219:8,14	presented 249:9
371:10	186:9 343:1	practical 12:17	221:21 223:9	president 60:12
portion 250:15	377:8	74:18 76:18,22	predication	128:17 132:7
portion 2 30.13 portrait 47:11	post 70:22	77:5 92:10	219:1,10	presidents 23:15
pose 6:12,17	130:19 197:1	106:5 113:13	predict 36:8	50:14 231:5
41:16,18	363:12	113:19 115:16	37:20 145:20	presiding 4:11
159:15 181:14	postal 106:20	159:4 171:15	183:3,21	press 31:9
185:5 255:22	107:13	235:7 292:20	184:19	131:12
368:16	postchurch	358:11,14	prediction	pressured 13:15
poses 6:11	85:13	practically	182:11	presuppose
position 70:4	postcollection	14:19 15:5	predictions	283:19
position /0.4	postconcetton	17.17 13.3	Predictions	203.17

pretend 197:8	297:8,16 300:4	24:14 25:2,14	126:15,17,21	217:5 218:5
pretext 44:18	328:1 332:16	25:20 26:6,7	126:22 127:10	220:12 221:1
pretty 77:2	principled	26:11,15,19	127:11 129:1	221:10 222:12
90:20 113:17	237:5 268:3	27:3,8,13,20	130:7 131:17	223:17,19
159:11 162:3	principles 61:9	27:22 28:2,6	132:2 133:14	224:10,12,15
193:3 194:20	61:10 64:10	28:11,16,17,19	133:16,19	224:21 225:2,6
195:17 199:4	84:12 86:17	29:3,10,20	134:1 135:10	225:12,16,21
343:22 381:16	100:2 134:4,6	32:15,16,20	142:18 143:13	226:2,7,9
prevent 126:14	164:12 180:1	33:16,21 36:6	143:17 144:7,8	227:2,19 229:5
131:8 174:17	192:18 204:9	36:17 40:13	144:15 147:5,6	229:15 230:12
175:1	231:21 234:22	42:1,5 43:3,19	147:9,12 148:1	230:13,16
prevention 85:3	272:1,17,20	45:8,13,16,16	148:14 149:4,9	231:2,3,10,13
previosly 229:2	273:3 274:18	46:8,12 47:18	149:11,14,20	231:14,18,22
previous 119:10	276:10 288:12	48:1,5,17 49:9	150:1,3,9,15	232:5,9,18,20
155:3 171:13	296:9,21 303:4	50:1 52:7,15	150:21 151:3	233:5,10,11,17
179:9,13	317:3 332:9	52:20 53:22	151:19 152:15	233:18 234:6,7
252:22 276:20	princples 220:5	54:9,13 55:2	152:21 153:5,9	234:9,14,16,18
278:17 286:13	prior 45:14	58:2 59:14	153:22 154:18	234:22,22
287:4 319:11	184:3 199:15	61:5,12,17	154:20,21	235:8,13,18,21
363:6	206:14 225:22	65:2 67:7 69:1	155:5,11,16,16	236:18 237:6,8
previously 90:5	264:11	69:17,18 72:2	156:11 157:16	237:11 238:10
117:6,17,20	priorities 227:8	72:11 73:6,8	166:19,22	238:12 239:12
148:5 228:2,11	priority 200:7	74:16,17 75:4	167:3,9,12,21	239:22 240:2,3
303:11	201:15 227:10	75:10,18 79:3	171:18,21,22	240:8,19
price 336:13,16	prison 104:6	79:4 83:7	172:2,4,10,18	242:16,21
pride 303:19	137:1	86:13 87:21	173:4,9 174:4	243:9,19
primarily 111:6	privacy 1:3,5	96:21 97:6,15	176:7,20	244:11,13,18
119:4 121:9	2:10 3:2,4,10	98:1 99:2	177:17 190:4	246:12,14
202:8 229:16	3:10 4:3,4,21	101:15,20	191:20 196:20	247:3,6,18,22
326:17	5:14,19,21,22	106:19 107:6	197:7,15 198:7	250:5,7 252:1
primary 94:18	6:4,4 7:22 8:2	109:11 111:4	198:20 199:2,6	255:18 256:2,5
256:2 324:19	8:4,10,13,19	111:13 112:19	199:8,17,20	256:8 257:1,11
346:9	8:22 9:5 10:15	113:7 114:5,18	200:8,10,14,14	258:12,18
prime 166:12	10:16,19,22	114:21 115:15	203:2,5,6,22	259:16 261:12
princeton 2:12	11:5,9,15 12:9	115:19 116:10	204:1,4,8,10	264:4,9,18
33:12,13	12:11,13,15,16	116:12 117:4,7	204:21 205:2,6	265:10 267:16
principal 94:18	13:3 14:4,7,10	117:12 118:5	205:10 206:11	267:20 269:5
215:12	15:10,11,19	118:12,13,16	206:16,19,21	269:22 270:4
principally	16:22 17:6,8	118:22 119:5,8	207:16 208:2,3	270:11 274:3
56:13 59:3	18:7,9,13,14	119:18,22	209:2,13,19	275:7 276:4
principle 23:3	18:14,15,17,18	120:12 121:1,9	211:2 215:10	277:6,7,15
61:20 62:7,18	18:19 19:9,11	121:14 124:2,6	215:13,18,20	281:1,4,10,11
86:21 202:14	19:13,15,16,22	124:11,14,19	215:22 216:4,5	282:6 284:13
221:6 254:1	20:3,6 22:22	125:4,4,6,19	216:13,16	285:5,14,15
	1			

287:5,18 289:6	224:13 225:18	proactive	51:17 72:17	procurement
290:5,7 291:1	private 3:17 5:2	218:22 267:22	76:15 103:3	250:5,9,11,12
291:3 293:22	5:7 6:7 14:17	proactively	104:12,15,18	252:7
295:1,20 296:2	27:17,19 42:17	205:9 322:2	104:19 105:4,5	procurment
296:6,12,19	42:19 43:5,16	probable 29:16	105:6,7 143:7	251:16
297:4 298:16	43:22 44:6,8,8	53:17 65:3,14	189:15 193:3	produce 164:4
298:16 299:3	44:16 45:7	73:1 81:21	203:19 256:19	246:11
299:13 302:14	63:1,13 102:11	169:5 362:3	311:9 344:7	produced 74:2
304:15 305:1	102:14,15	probably 27:14	376:16	produces 16:5,6
306:3,10	103:22 104:4	58:15 101:5	procedural 28:1	product 26:17
310:11 311:18	104:18,21	120:16 137:6	procedures	35:8 126:14
312:2,4 313:20	124:20 125:12	191:17 276:12	114:5 177:13	150:22
317:10,11	125:13 127:16	306:13,18	300:5 314:22	products 111:19
330:1 332:2,4	127:17 133:19	307:10 317:16	376:2	113:2 125:10
332:7,21 333:1	133:21 141:7	318:17 337:1	proceed 4:20	125:18 131:20
333:3,17,18,19	153:21 141:7	342:22 343:2	134:19	150:18 154:20
333:22 334:2	186:9,18	344:13 360:10	proceedings 4:1	professional
334:12,22	216:10 229:7	probe 11:17	130:11 385:6,8	140:17 242:16
335:19 336:1,8	231:16 233:12	probe 11.17 probing 41:3,5	,	247:20
337:2 338:17	231:10 233:12	_	process 56:9 117:18 130:13	professionals
		problem 13:14 21:11 27:7		190:12
339:20,22	264:7,8,19		146:5 174:9	
340:5,11,17,17	267:15 284:19	47:17,22 48:8	191:8 218:8	professor 18:1,3
340:18 341:2	285:4,20	48:10 49:20	223:7 231:14	33:11 45:10
341:12 342:2	287:20 288:14	51:18 62:17	232:15,17,20	54:8 89:19
343:6,7,15,18	290:14 291:14	92:1,7 117:12	233:6 236:9,10	110:12 142:8
344:5 345:9	293:3 294:14	117:13 118:4	240:18 250:6	169:17 186:22
348:15,17	296:10 297:6	120:11 121:17	251:16 274:2	194:7 286:21
349:9 352:21	297:14,17,19	137:11 146:18	277:5 283:14	287:2,2,6
353:7 356:21	297:21 298:2,7	171:8 264:21	308:18 316:8	296:14 305:22
357:6 358:21	298:10,14	274:20 277:3	330:20 344:16	326:9 328:2
362:4 363:7,11	299:10,16,21	305:2 309:14	344:18,21	335:22 341:17
363:20 364:3	302:13 303:5	316:21 320:1	345:20 346:11	374:4 375:17
364:20 365:3	314:5 317:15	335:6 346:5	347:8 356:9	381:9
365:19 366:18	317:21 327:21	353:13,15	382:22	professors 92:17
368:13,20	328:7 329:11	356:7 371:11	processes 29:21	195:12
369:2,7,12,15	331:4 336:5,8	371:16	57:21 96:2	profile 75:14
369:21 370:7	341:20,22	problematic	117:21 126:9	148:9 268:16
370:18 371:2	342:5 343:12	361:12	127:3 132:18	profiles 159:10
375:2 376:1,3	345:2 352:17	problems 20:16	173:22 215:4	162:12
376:11,12	354:9 356:13	20:20,21 21:2	224:6 252:13	profound
383:19	357:4 367:17	46:12,13,15,16	330:22	151:20
privacypreser	379:18	46:17,20 47:1	processing 46:4	program 7:6
40:18 106:12	privilege 287:15	47:2 48:7,8	323:2 324:11	10:8 25:5
privacyrelated	325:4	50:5,5 51:17	325:5	30:17 31:3

56.10.04.14	mmahihitad	101:17 111:2	52.17.05.10	120.22 191.2
56:18 84:14 90:1,9 92:4	prohibited 122:20	113:21 116:11	52:17 85:10 86:13 111:3	129:22 181:2 202:1
ll '	· -	118:16 121:14	132:8 163:5	provides 100:9
126:21 127:2,8	prohibitions 128:19		166:1 204:19	113:13 216:2
		124:6,8 133:6		
	project 311:12	167:14 199:5	208:2 221:1	221:11 225:14
11	prominent	204:11 208:3	225:17,21	226:18
163:17 174:3	122:4	228:9 230:16	229:15 232:2	providing 24:2
II	promise 99:4	233:9 237:7	233:7,14	83:8 85:10
180:20 181:1	122:10	258:12 264:10	234:17 280:22	104:8 111:3
II I '	promised	264:12 289:5	281:1 282:8	115:10 135:8
204:13 206:20	156:16	330:1,22 333:2	296:16 333:1	207:12 209:6
	promote 154:21	333:22 334:3	341:9	210:12 211:8
225:6 229:21	234:9	369:1 376:11	protective	227:15 249:12
II ' ' I '	proof 225:1	protected 31:10	299:12 332:7	266:7 272:5
II	proper 23:14	42:11 114:1	protects 12:5	280:4
257:15 260:14	48:18 74:15	126:4 133:17	20:7 125:18	provisions
261:21 275:7	210:3 315:3	150:8 167:17	167:15 207:16	122:14 365:6
311:2,17 332:4	proportion	191:14 211:4	209:2 217:5	proxy 274:10
332:6,11	317:13	222:11 264:5	304:15 348:17	pseudoprivate
359:18 360:18	proportionality	protecting 28:16	protocols 51:8	107:15
365:9,22	317:6 345:4	52:7 61:11	proud 28:12	psychologists
366:14 372:1	362:14 366:3	129:1 133:19	prove 113:9	122:22
programs 30:15	proposal 212:9	142:18 234:5	proved 134:5	public 1:12 4:4
32:2,7 46:3	proposals	242:21 284:13	proven 235:2	5:4,7 7:10 14:1
56:16 58:14	358:17	287:19 289:5	366:18	14:4 15:2
64:6 81:6,18	proposed	306:3 341:12	provenance	33:12 48:4
128:7 129:3,7	354:11	341:14 367:7,9	326:4,5 365:17	66:2,2,5 67:16
130:5 168:14	proposes 358:8	protection 42:4	provide 24:12	74:2 75:2,4
	proposition	46:2 113:14	84:16 105:13	84:8 96:14
200:21 201:2	96:15 233:2	114:5 154:3	112:17 130:16	115:5,11 120:1
204:15,18	proscribed	168:20,21	135:5,20	129:2 130:22
205:4,7,12	334:20 365:11	169:4 194:20	168:20 169:2,3	148:14 163:14
· · ·	proscriptions	194:22 200:9	210:9 211:11	180:7,8 197:17
215:18 216:2	324:8	200:11 206:5	211:22 213:14	207:8 211:10
	prosecute 16:1	216:14 229:20	214:1,15 218:1	213:3,6 255:5
II '	prosecutor	258:16 290:7	239:15 276:2	278:21 284:8
245:16 247:14	70:15	290:11,12,17	294:19 295:5	287:20 288:15
	prospect 95:12	296:2,7,20	301:9 302:3	291:14 292:18
II 1 '	prospective	297:4 299:4	305:18 370:8	298:15 317:20
331:21 333:9	267:19	306:10 332:2	provided 43:4	328:5 329:12
	protect 9:1 20:2	349:14 362:10	175:13 219:5	330:5 336:5,18
progress 153:13	20:3,6 24:3	376:12	provider 182:16	337:2 339:21
154:7,14	28:20 29:4,20	protections	188:4,20 189:5	340:1 381:7
,	· ·	-	· ·	
234:21 284:8	29:22 31:11	24:14 43:4	providers	385:4,17

publically 219:4	pursued 260:20	170:1 171:12	91:16 100:9,20	337:8
publicity 5:8	260:21 280:3	174:1 175:9,19	100:22 101:1	quorum 4:13
publicly 14:19	pursuing 213:18	179:5,13	102:3 109:1	quote 8:7
15:3 75:11	push 96:5	181:12,20	110:5 133:22	136:20
92:3 208:12	201:22 235:20	182:13 183:8	134:13 141:1	
339:1,13	269:3,13	185:4,5 186:14	141:15,21	R
publish 129:17	put 7:7 17:4	192:2,5 195:1	150:7 159:16	race 68:18
published 82:5	27:6 39:8	196:18 197:9	165:5 175:21	rachel 2:4 4:13
publishing	52:20 58:22	197:10 199:9	181:9,14 185:8	80:14 159:18
271:13	83:2 93:22	199:15 200:6	194:6 206:1,3	271:4 285:7
pull 203:17	144:12 150:3	212:16 238:9	215:6 230:6	racial 122:1
316:7 323:21	150:12 152:11	238:16 239:11	232:5,7,10,15	radically 148:9
345:21	160:15 163:6	239:18 241:5	237:19 238:11	raise 110:9
pulled 60:16	167:1 169:9	241:12 252:15	238:20 239:3,9	276:4 305:4
pulling 58:16	172:22 201:15	255:13,22	248:5 255:2,4	357:7
purchased 37:6	229:11 242:3	257:2 258:11	260:15 262:12	raised 74:15
purchases 35:8	269:14,18	262:9 267:14	262:14 263:2	326:9
purporting	280:5,8 370:21	274:12 276:13	274:5,15,19	raises 357:1
271:9	377:15	276:15 279:5	275:10 277:9	ramification
purpose 69:9	puts 25:18 38:20	281:3 282:17	278:3,6,8,12	17:5 140:12
94:9 97:18	245:18	283:4 286:18	278:15 280:3	ramifications
118:9 256:12	putting 60:6	289:4 293:5	282:20 284:16	13:2 137:12
260:17,18,19	151:8 268:5	313:16 318:10	286:9,11,12,14	139:14 202:4
261:1 265:1,1		318:12,15,20	312:11 318:6	range 11:10
272:11 275:2	Q	321:20 326:10	322:5 343:16	101:20 152:7
288:21 289:2	quantify 164:8	334:9 335:18	357:7 368:9	172:13 253:8
296:16 300:16	196:21	344:11 348:7	381:7 383:6	319:1 383:11
300:18,21,22	quasifederal	357:22 358:10	quick 66:12	ranging 237:10
329:6,7 359:19	106:19	358:16 360:14	71:3 76:17	rank 319:19
359:20 360:1	queries 164:10	361:11 363:5	78:3 175:9	rap 14:13,15,17
362:16 366:19	querying 60:13	363:22 367:14	179:12 183:19	rapid 226:15
purposeful	question 9:15	368:16 370:11	198:18 203:12	rapidly 190:6
178:11	10:1 26:9	372:19 373:5	313:15	rare 12:5 295:3
purposes 79:20	31:16 43:9	374:3 379:4	quickly 51:4	rarely 133:12
93:13 112:2	55:7,8,11 61:8	questionable	53:3 146:6	rat 13:13
186:19 273:5	63:19 66:19	112:22	175:22 198:11	rate 57:1 58:16
281:20 299:9	74:15 75:3	questioning	200:12 267:14	315:21
318:4 352:22	76:9 82:20	74:11 110:3	277:14 354:7	rates 315:22
365:15	90:16 96:2	141:17	quite 12:19 17:7	ratings 184:17
pursuade 243:6	98:11 102:18	questions 6:11	70:11 74:4	ravaged 137:19
pursuant	105:19 107:21	6:12,14,16	87:12 151:20	reach 91:11
204:18 278:20	110:8 142:1	9:17,20 10:2	224:3 240:5	249:17
pursue 57:14	154:16 155:14	41:1,3,6,14,17	246:10 249:16	reaching 33:6
221:9	159:22 165:16	81:16 88:8,10	252:8 318:3	reaction 353:2
		<u> </u>	<u> </u>	l

l 				1
reactions 360:11	157:20 158:13	reasonable 23:1	recognition 42:7	175:12 203:3
reactive 267:19	158:13 159:21	60:12 65:10	55:2 148:3	284:21 306:21
read 31:4 45:19	161:13,17	70:4 111:11	362:7	385:8
58:7 83:6	163:10 170:3	147:5 148:13	recognizable	recorded 7:6
181:13 339:10	177:3,3,6,15	157:20 162:7	54:10	34:9 163:20
351:17	177:16 178:5	169:4 198:7	recognize 8:22	330:10 385:7
readily 75:5	187:3,4,5,6	292:17 334:11	54:11 64:11	recording
160:19 273:22	190:1 195:21	362:2 363:7,10	68:1 75:18	329:15
reading 103:4	196:4,4 197:4	363:19 364:2	123:6,18 125:7	recordings
162:22 220:9	197:5,10,10	366:15,17	131:18 153:9	34:11
241:9 301:19	200:13 220:5	368:12,13,20	155:3 231:5	records 14:11
339:15,16	225:4 239:4	369:2,6,11,15	270:16 284:5	14:16 15:16,16
reads 324:15	248:11 250:7	369:20 370:6	327:18	98:21 99:5
real 77:10 97:6	251:16 252:7,8	reasonableness	recognized	120:7 163:3,17
97:6,7 113:13	256:21 259:1	65:16 70:17	42:11 65:5	168:1 221:12
132:3 184:4	259:10,17	377:8	133:13 151:13	221:14 222:6
252:21 257:1	263:3,22 270:3	reasonably 59:2	217:12	278:19 301:14
266:21 292:3	271:10,18	69:20 106:4	recommend	301:17
328:21 331:2	272:4,8 275:9	155:2 177:18	61:11 93:13	red 2:14 17:21
379:12,22	277:12 282:1,2	182:11	303:1	22:14 25:5
realistic 221:7	282:21 287:16	reasoned 327:9	recommendat	110:2 207:6
realities 99:15	287:17,21	reasons 79:1	92:15 213:19	286:4,6 354:4
reality 77:10	288:15 289:3,6	264:2 268:4	240:12 373:9	redacting
84:15 97:21	291:4 292:15	292:20 299:15	recommendat	307:12
realize 33:1	294:18 307:6	319:6	92:11 128:15	redactions
121:21 148:12	308:9 312:12	reassert 68:2	171:16 239:15	220:11
156:22 167:9	313:14 316:2	reassociate	279:19 288:1	redefine 118:12
357:1	323:14 326:14	196:14	300:1	redress 83:15
realized 12:16	328:7,8 329:13	reassurance	recommended	282:9 294:11
210:16 347:12	330:19 332:21	211:11	163:22	294:15 295:4,9
really 13:20	336:6 346:13	reassure 120:17	recommends	reduce 40:11
25:13 27:9	357:16 360:14	267:2	93:16	174:15
32:6 47:6 48:7	373:6 374:16	reassures 56:9	recommit 303:2	reduced 174:22
53:9,10 55:4	378:20 381:5	rebecca 3:14	reconcile 55:1	reducing 211:7
58:10 67:9	reason 15:17	recall 70:21	157:13	redweld 306:14
68:18 71:3	25:17 53:9,11	321:5	reconciled	307:1
72:20 76:10	69:14 82:4	receive 5:18	315:13 316:1	reevaluate
84:3 88:21	121:8 155:9	88:6 136:7	reconstruct	134:3
89:20 93:6	175:6 247:8	332:19	39:18 184:10	refer 355:5
107:21 116:3,6	291:5,12 292:1	received 128:9	reconstructing	reference 75:6
120:16 143:3,5	292:15,22	129:13,19,22	179:5	referenced
143:6 145:7,20	293:1,2,2,3,4	139:10	reconvene 203:1	74:16
148:22 150:4	339:7 349:3	receptacle 8:14	record 59:20,21	referendum
155:22 157:1	362:5,17 372:3	recipient 139:22	97:9 109:4	361:1
		<u> </u>	<u> </u>	I

				1
referred 126:17	regulates 103:1	releasing 14:9	remedies 281:10	177:4 206:11
186:4 192:3	regulating 24:11	164:9 201:20	remedy 282:6	253:20 302:13
referring 68:4	38:14	relevant 28:3	remember	repository 21:14
185:21 326:17	regulation 42:19	100:12 125:22	106:21 119:18	49:16
refers 97:12	43:7 102:13,19	134:7 203:14	135:21 136:3	represent 45:15
refine 232:15	103:11,12,14	206:15 289:1	remind 88:9	197:14
reflect 97:9	103:19 108:6,7	301:4 307:4	109:19	representation
170:21 355:15	108:12 290:17	352:7,8 370:16	reminded	364:16 380:7
reflected 174:10	regulations 7:11	relied 375:6	203:20 214:21	representative
229:17	384:5	relies 121:9	reminder 255:3	32:21 253:12
reflects 29:13	regulatory	religion 11:3,16	318:7	represents
382:5	372:18	11:17,19,21	reminders	155:7
reform 127:22	rehash 354:3	12:2,3 26:16	214:16	republican
130:15 168:5	360:15	44:10	reminds 159:2	69:22
179:21 212:3	reidentified	relinquish 13:22	remotely 197:14	repurpose
380:16	113:11 114:2	relocate 122:16	render 214:7	382:15
reforms 130:3	reidentify	rely 74:18 124:3	renee 7:15	repurposing
refrained 16:14	160:18	190:9 194:1	109:22	322:21
regard 33:3	reigning 44:15	246:10	renting 321:7	requested
126:1 128:1	reinforcing	remain 45:8	repealed 123:12	365:15
132:6 219:8	184:3	119:7 122:12	repeat 198:8	requests 167:21
220:8 221:19	reiterate 341:9	125:12,21	repeatable	168:14 302:5,8
222:9 246:4,13	relate 47:2	228:6	232:16	320:22
281:22 285:6	183:4 285:5	remaining 86:8	repeatedly	require 30:22
294:15 317:17	related 49:9	101:12 154:15	298:15	59:4 168:5
regarding 5:2	124:20 128:13	223:22 274:18	replace 101:9	218:22 219:1
6:4 87:4	232:7 241:4	remains 44:5	350:5	219:14 228:8
204:12 213:20	295:17 344:7,8	remark 278:17	report 50:14	370:21 371:4
272:6 278:21	relation 21:21	340:10	74:3 82:5	required 102:13
301:10	relations 115:11	remarkable	84:11 119:12	168:10 172:12
regardless	relationship	375:9	162:22 179:20	208:3 272:16
118:22 168:6	63:17 167:16	remarkably	184:17 213:20	298:22 346:21
231:8 303:5	239:14 302:20	383:7	223:14,15	requirement
365:1	337:22,22	remarked 8:4	256:9 279:19	29:17 71:21
regards 220:14	338:1	373:15	351:17,19	202:5 311:16
regime 230:11	relationships	remarks 7:2	358:1 362:19	314:10
register 4:10	54:20 153:3	45:14 109:20	reported 1:22	requirements
regnant 86:8	242:10 263:18	140:22 141:14	130:20	174:4,5 227:8
regular 224:11	265:13	157:22 172:6	reporting 130:2	239:16 272:19
240:17 373:19	relatively 76:2	217:1 237:21	220:16 240:10	327:6 343:7
regularly	225:13 298:3	281:12,14	302:6 336:12	requires 31:18
133:22	release 155:20	284:15 286:2	346:21	52:11 91:4
regulate 46:19	released 157:8	remedial 281:5	reports 31:5	130:12 131:4
78:22 211:16	222:15	281:8	57:16 163:19	188:19 219:9

requiring	70:8 127:3	retailer 35:7	202:14	161:7 173:9
111:18 112:17	227:7 253:22	retain 62:1	revoke 201:22	175:3 176:17
370:14 371:12	262:9 305:2	118:2	202:16,18	178:7 181:7
research 18:3	355:4 357:21	retained 126:2	rewriting	186:14,21
39:15 92:2	responses	221:4 222:4	366:22	195:21 196:2
144:4 234:5	165:17	300:17	rhetoric 294:2	193.21 190.2
250:15 251:10			richards 3:14	201:19 202:15
250:13 251:10	responsibilities 245:10	retaining 134:5 263:20		201:19 202:13
			225:11 226:4	
349:19	responsibility	retention 40:20	238:9 250:10	241:19 242:22
researchers	240:13 244:1	90:4 93:2	262:10 268:22	243:1 247:15
144:3 160:16	responsible	256:10 257:22	271:11 272:9	248:5 251:2,3
reservation	18:20 19:1	258:4 259:3	273:10 277:1	251:13 252:8
321:8,9	209:15 236:22	322:6,11 351:9	283:6 354:20	253:5 258:8
reshapes 226:11	responsive	351:10 371:8	rid 52:8	259:12 261:18
reside 231:9	298:19	retrospective	ridiculous	262:7 263:12
resides 156:2	rest 88:16 190:6	347:1	197:19	263:17 268:6
resilient 111:16	293:16 327:12	return 195:6	rifle 307:2	275:10 285:9
resist 118:18	327:16	returned 222:1	rigged 200:15	293:16 310:19
121:9	restore 128:5	returns 192:13	right 4:21 5:1	314:2 315:3,6
resisting 377:4	restrain 29:22	319:12	17:22 19:16,17	326:3 333:19
resolution	restraining	reveal 76:14	26:17 48:14,17	335:15 338:15
170:15,17	27:20	151:15 153:2	51:5,5,14,16	344:16 345:12
resources 68:17	restraints 300:9	revelation 47:13	51:20 53:2	347:6 350:19
150:2 160:19	restricions	revelations	54:4,10,15,15	350:20 361:19
194:4 245:13	245:20	128:6 131:15	57:11 60:5	362:1 364:16
248:7 294:22	restricted	reverse 159:17	65:2 66:22	364:21 365:10
310:3,8 319:8	300:13	194:15	72:10,12,18	365:20 367:14
343:2	restriction	review 30:9 31:1	76:6 77:9 82:8	378:9 379:13
respect 45:3	42:16,17	57:20 59:9	82:12 83:14	380:5
90:13 97:17	121:13 122:12	60:3 65:19	84:5 85:16	rights 8:12 11:2
115:11 179:6	245:19 301:2	71:1 74:1	87:1 88:20	19:2 51:7
193:2 198:2	376:20	116:1 130:12	90:15 96:21	83:15 84:20
231:8 334:10	restrictions	168:13 221:11	97:1,17,18	99:2 130:8
361:11 365:21	81:10 118:17	221:13 223:6,8	99:20 107:18	131:17,22
380:3,5	121:10 124:3	223:17 251:2	117:12 118:10	150:15,21
respectful	245:18 341:18	253:18 281:20	118:22 124:9	157:6 191:22
189:15	result 176:9	371:1	129:12 132:3	202:1 226:21
respond 158:22	187:19 228:21	reviewed 358:3	134:1 136:4	234:11 295:9
348:19 351:12	266:18 315:20	reviewing 30:5	138:10 139:7	329:19 378:10
responded	379:3	reviews 30:3	140:16 146:9	riley 133:14
298:15	results 177:4	173:4 216:1	146:10,11,20	151:13
respondents	290:2 330:8	223:5 281:16	147:14 153:1,1	ripe 59:21
16:10		revocation	153:10 157:4	ripe 39:21 rise 114:22
	resume 109:3,6 284:18	201:18 202:3	158:17 160:13	risk 13:17 40:11
response 57:9	204.10	201.10 202.3	130.17 100:13	118K 13.1/40:11

49:15 60:18	73:21,21 74:6	rubber 259:11	132:12 133:2	353:14,17
62:5 92:19	74:7,9 83:13	rubric 9:5	164:5 230:2	354:14 381:13
93:1 106:6	116:20 207:12	rule 38:19 99:10	278:22 279:2,3	says 72:9,12
111:12 114:14	225:13 239:2,5	126:7 131:6	279:21 282:9	96:4 163:1
115:12 119:16	239:18 247:3	135:2 159:16	284:6 289:9	314:16 332:21
159:6,10	305:10 317:1,2	315:5 371:22	292:4 301:8,8	342:3 347:1
162:12 163:4	372:13 373:7	375:12	302:10,11	349:9 352:21
183:20 235:8	374:6,8,9	ruleladen	343:17 369:7	353:8 357:12
235:17 236:9	375:18 376:22	325:14	sadly 57:4	357:14 362:8
236:10 258:18	377:18 378:14	rules 28:6 38:11	safe 133:1	381:10
259:16 262:3	379:5 380:3,10	95:1,8 113:7	safeguards 87:4	scale 50:17
263:2 275:6	rolebased 127:5	132:19 135:4,9	safer 40:10	52:20 73:15
280:5 290:18	172:14	135:12 140:15	safety 40:6	152:18 169:2
290:18,18,22	roles 12:3 127:1	157:21 158:2,3	133:7,8	248:2 260:5
291:3,12,21,22	127:7 172:17	157.21 138.2,3	sake 20:7,8 27:3	261:6 302:18
330:21,22	246:2 355:12	159:11,13	192:14	318:3,5,5
331:16 334:6	roll 288:11	168:12 169:11	sales 127:6	334:8 344:7
337:11 350:10	room 22:11 60:9	169:15,21,21	172:16	349:17
381:15 382:1	78:22 115:22	171:6,21	salient 27:15	scared 373:5
risks 43:19	116:4 137:7	171:0,21	salt 27:6	scenarios 86:11
45:16 49:11	144:22 165:12	190:8 192:16	san 27.0 sam 286:2	167:10
58:2 61:5 93:4	193:14 201:14	210:22 211:17	sam 200.2 sample 184:2	schedule 222:5
104:2,9 162:13	304:21 329:14	211:20,22	sample 184.2 samples 135:20	schemes 281:8
225:2 231:22	rosenzweig 2:14	211.20,22	samples 133.20 sampling 164:4	scholar 115:15
232:18,22	25:4,9 33:9	213:5,8,9,12	sampling 104.4 sanchez 66:16	scholar 113.13 scholars 113:4,8
233:5 236:1	50:8,11 53:20	213:13,13	sanctions 95:7	113:12
237:7 256:8	55:6 58:5 69:5	238:18 335:13	sanctions 33.7 sand 300:10	scholarship
258:10 261:5,7	76:17 84:10	343:18 344:9	sat 136:20	36:17
261:8 262:1,2	94:14 98:5	348:6,12 356:6	satified 210:2	school 3:18 18:4
266:21 275:4	99:7,18 100:21	rulings 130:12	satisfied 43:3	110:13 117:6
291:6 343:12	101:11 106:17	run 37:22 55:22	177:19 187:15	161:12 196:1
risky 235:21	107:7	126:19 188:21	satisfy 157:21	287:3 292:5
270:4	rough 163:15	336:2 360:10	316:5	science 33:12
road 259:11	round 110:3	running 130:21	saw 164:1 278:7	77:11 110:15
roberts 133:13	141:17 146:12	182:7 200:8	300:11 375:1	233:17 234:7
robust 84:20	route 245:5	343:10	saying 47:6	234:17 261:15
125:15 210:5	routine 222:15	russians 189:13	150:20 160:2	313:11 316:8
211:2 252:20	222:17	1 ussians 107.13	163:19 167:1,5	335:15,16
253:14 297:4	routinely 123:1	S	167:8 178:20	sciences 181:16
303:2	229:10	s 3:13 36:4	178:22 187:8	scientific 235:1
robustly 299:16	row 9:11 109:17	127:19,20	260:8 284:2	236:9,15 237:5
rogue 346:19	109:22 116:16	128:4 129:10	298:10 310:20	scientist 33:19
365:18	185:5,6 186:21	129:14 131:15	310:22 348:2	134:18
role 32:14 65:20	207:6 286:3,20	131:17 132:1,9	350:10 351:15	scientists 237:11
1016 32.14 03.20	207.0 200.3,20	101.11, 102.11,9	330.10 331.13	Scienusts 23/.11
<u> </u>				

237:13	76:11 105:21	304:9 310:20	133:20 134:15	145:5 148:20
scifs 253:18	secrecy 32:1	314:5 317:16	135:7,9,13,14	148:21,22
scool 2:15	192:8 208:10	317:21 327:21	140:9,16,18	150:2 155:14
scope 164:1,8	208:21 210:21	328:8 329:12	143:13 155:19	160:17 177:2,9
301:3 302:4	373:8	329:12 330:6	156:21 162:17	183:22 190:20
334:8	secret 66:4	331:4 336:5,6	166:13 176:13	198:1,9 199:4
scratch 53:19	98:18 214:18	336:8,18 337:2	176:19 188:8,9	223:10 236:22
65:8	302:2 355:17	341:20 343:12	188:11,15	245:5 249:10
screen 174:9	355:18	345:2 352:17	190:3,12	256:12 266:11
screening 59:16	secretary 25:7	356:13 357:4	200:11 204:5	280:16 290:15
84:19	secretly 301:15	367:17,17	218:16 219:19	294:13 307:1,3
scrutiny 29:12	secrets 18:8	379:18	223:2,4,5	307:7 308:6,12
60:16	19:7 32:10	sectors 284:19	225:3,12,22	320:14,16
seal 385:12	115:3 208:11	287:20 342:6	226:3,15	341:1,3 342:10
search 42:3 52:9	214:11,17	secure 12:21,22	229:10 246:19	372:11 376:20
65:10 198:13	267:9	18:20 49:10,11	256:21 257:1	seeing 57:8
198:14 300:12	section 90:1,9	72:10,12 94:7	257:10 272:12	158:18,18,19
searched 91:20	90:14 92:3	113:2 138:4	280:1 285:7	188:2 290:14
searches 52:10	103:5 119:13	143:21 155:16	290:4 291:2	297:8 338:16
53:8,18 70:10	128:16 175:10	209:21 253:17	293:2 294:10	370:12 376:16
72:10,13,15	278:20 301:4	secured 256:17	294:19 296:3	seek 125:14
198:14 290:5	301:11,15	security 3:7,14	297:20,22	131:2,5,20
314:20 369:1,4	330:11 351:17	8:18 10:8	298:8,18 299:8	222:8 365:5,6
searching	369:5	14:11 19:15,17	302:5,7 303:12	seeking 132:12
192:22	sector 3:17 6:7	19:21 23:1	303:14 310:20	241:14
second 5:22	43:5 104:18,21	24:15 25:3,8	313:20 321:15	seemingly 35:12
13:22 19:12	124:20 127:16	29:14 42:2	327:11 330:20	36:14 89:5
34:2,13 37:14	127:17 133:20	46:3 49:8 54:1	331:3,12 333:9	seen 19:16,17
70:1 109:10	133:21 174:4	61:18 64:6	335:12,19	44:12,13 95:18
111:10 115:2	186:9,19	66:8,13,18,20	338:21 339:7	109:22 129:4
116:16 117:2	200:16 216:10	67:4,5,6,9	343:11 345:8	337:4 374:21
119:2 128:21	230:18 231:16	69:11 73:6,12	367:1,7,9	sees 18:15
199:10 202:22	233:12 234:12	73:13,14,17,18	375:19 381:21	segment 198:3,4
238:1 269:11	250:4 264:7,8	81:1,12 83:18	382:13 383:19	segments 198:2
285:16 290:13	264:19 267:15	84:13 85:10	see 6:6 19:9 24:1	segue 159:21
297:18 301:9	285:4,20	89:9,13 100:5	24:2 27:12,21	seinfeld 321:7
324:14 332:12	288:15 290:15	104:6,10	49:8 50:1,21	seized 290:9
334:21 346:11	291:14 293:3	111:20,22	55:7,8 56:6	292:8,10
364:18 381:15	294:14 296:11	112:12,18,20	57:4 59:14	seizure 42:3
secondary 46:4	297:6,15,17,19	112:21 114:20	69:14 76:7	seizures 72:11
58:17 84:18	297:21 298:2,7	115:20 116:13	86:8 95:16	72:13 369:2,4
secondly 186:15	298:10,14	123:22 126:4	104:20 116:6	select 174:7
319:10 373:4	299:10,16,21	129:14,18	135:13 142:14	selected 226:6
seconds 32:13	302:13 303:5	132:22 133:7	144:10,13,22	selfcorrecting
			•	•

158:20	separately 62:13	203:7	shes 70:8	sign 126:15
selfevident	sequential	set 108:10	shift 371:2	167:1
335:3	235:10	113:10 125:15	shifting 213:4	signal 355:21
selfhelp 23:22	serendipitous	163:2 222:5	213:11 251:21	signals 231:6
68:5,19	50:18	231:12 232:10	259:3 330:1	320:2
sell 131:20	series 8:20	264:21 271:22	shipping 126:14	signant 227:22
semiclear	29:20 198:1	299:22 316:21	shocking 163:13	signed 51:9,14
158:11	206:1	322:9,10	shoes 361:9	354:15,16
senate 117:7	serious 193:3	374:10	shop 238:8	significance
164:1 179:16	194:5	sets 117:16,19	239:13	54:12 98:13,14
303:11	seriously 171:21	167:15 323:16	short 86:1	185:13
senators 193:8	177:20	325:17 335:1,4	117:21 137:18	significant
193:12	serve 361:10	359:21	344:15 361:1	20:15 43:19
send 9:21 136:7	served 204:15	setting 69:18	shortcuts 350:1	58:15 59:11
193:14,17	206:16 288:21	86:19 294:14	shot 67:18 99:13	216:19 218:1
sends 343:17	303:11	settings 114:3	shouldnt 18:10	223:6 228:21
senior 25:6	server 156:2	197:17	55:4 67:9	234:21 240:1
224:10 226:1	serves 12:13	settle 379:14	95:18 171:8	253:1,11 309:5
238:13 239:22	67:7	seven 9:10	180:2 311:22	314:16 338:4
295:15	service 26:21	109:20 137:14	361:16 375:7	361:3 367:13
sense 28:5 31:10	106:20 107:13	137:16 286:1	show 65:2 267:6	significantly
83:18,19	112:15 122:19	severely 377:2	267:6 283:15	240:5 248:18
105:19 151:6,7	123:4,13	shame 28:11	283:16	248:19 306:9
161:18 163:15	126:14 127:7	shannon 7:14	showing 16:9,13	silicon 135:1
166:21 221:11	129:22 139:22	shape 20:4	shown 37:12	311:12
244:1 246:6	150:22 152:9	226:20	177:5	silly 290:2
248:11 251:4	181:2 182:15	share 13:7,9	shows 11:10	silver 172:7
271:22 273:8	188:3,19 189:4	17:10,12,13,14	39:15 67:13	332:9 370:7
274:8 317:18	202:1 208:13	105:2 198:3	shredder 307:10	simialar 268:7
320:11 322:6	208:18 299:6	220:17 264:20	shut 283:2	similar 29:14
338:2 344:13	328:11	375:14	side 19:20,21	122:18 212:7
365:14	services 34:10	shared 221:2	58:14 66:11	239:10 264:20
sensitive 35:11	125:8,11,18	222:20 283:3	68:20 73:13	266:3 267:17
35:12 115:6	126:19 127:17	shares 13:5	130:13 142:12	278:21 302:13
120:20 168:8	131:10,21	sharing 17:16	142:13,13	320:5 344:6
268:11,16,19	140:1 150:18	119:20 123:11	155:6 170:2	355:12
sensitivity 152:2	151:11,17	142:5 149:17	216:16 219:17	similarities
172:10	155:12 158:4	241:12 325:19	219:18 221:16	264:12,15
sensor 17:3	188:19 199:3	sharon 7:14	222:10 261:6	similarity
sent 137:2	228:5 299:2	sharply 112:8	261:15 264:17	264:17
139:10 326:13	serving 122:20	sheer 37:19	267:5 277:22	similarly 37:10
sentence 105:17	123:12	228:19	371:17	simone 7:15
separate 8:1	session 2:9 3:1,9	sheet 14:13,15	sides 213:15	simple 136:17
167:6	3:16 202:22	14:17	256:21 382:2	137:9 329:1
	l		<u> </u>	<u> </u>

Tr.				
125.20 141.4	-1-:: 200.2	26.21.22.6	47.9.12.54.4	261.6 262.10
135:20 141:4	skipping 309:3	26:21 33:6	47:8,12 54:4	261:6 262:19
155:13	sky 136:22	155:10 289:18	55:5 60:3	263:1,3,6,14
simplest 124:5	skype 137:3	296:14	61:16 175:4	268:3 269:5,10
simply 28:10	sliding 169:2	society 8:15	186:4 201:10	269:12 270:14
62:18 84:12	slight 199:13	11:4 17:2	245:4 249:17	271:2 273:11
106:1 113:1	slightly 21:7	19:18,20 20:3	353:17	274:2 275:1,17
115:8 143:15	269:16 316:20	20:4,9 22:6	someones 156:2	277:9 283:12
156:20 175:14	sloppy 362:7	27:4 47:20	268:16	284:1 289:15
314:7 322:21	slow 121:21	50:19 54:13,16	someplace	290:2,21
323:12 345:8	123:18 379:20	54:22 55:3	275:16 292:11	294:16 306:1
364:10 370:17	small 51:21 59:7	56:10 57:7	somewhat 25:21	307:22 329:4
simultaneously	66:4 242:9	67:7 70:6	33:15 100:18	330:20 333:18
187:22	244:21 245:9	121:21 123:17	158:20 173:18	335:14 338:12
single 40:2 62:1	246:5 299:22	137:21 147:17	183:12 230:13	338:20 340:3
98:9 121:8	310:6,6 371:7	167:15 191:8	240:10 255:12	340:22 352:2
140:3,3 168:11	371:10 374:11	229:4 291:19	256:3	366:13 368:12
168:11 270:17	smaller 44:4	302:22 312:5	sophicated	369:10 378:3
sit 126:22	138:12,12	societys 20:8	280:7	379:3,20
272:18 316:16	228:8	societywide 17:1	sophisticated	sorts 91:21
sites 39:12	smarter 110:22	software 111:20	41:9 111:1	soul 192:22
sits 108:7	187:11,11	126:12 156:5	112:22 142:4	sound 136:22
sitting 6:22	smartphones	165:8 172:15	199:4	177:18 328:22
22:11 69:21	139:6	173:2 176:21	sorry 51:1	329:1
174:7 286:3	smile 8:6	177:12	193:16 194:17	sounds 88:2
306:14	smorgasbord	sold 24:6	201:11 207:3	197:20
situated 205:2	297:2 326:19	solely 124:2	255:8 259:5	source 323:13
situation 99:12	smudges 171:1	solid 380:7	262:6 263:10	sources 89:7
situations 8:20	snapchat 156:15	solitude 5:6	287:13 318:16	214:18 323:13
259:19 348:20	157:11	solove 2:15 18:2	sort 26:11 30:21	soybeans 193:15
six 16:13 45:16	snooping 24:4	18:5 45:11	55:13 57:8	space 5:2 138:3
177:2 273:16	snowden 16:9	46:7 51:21	60:3,11,19	304:7 338:15
274:18	16:14 57:9	52:1 71:3 76:5	76:19 81:15	349:22
sixteen 45:15	120:6 163:14	83:4 93:5	87:12,13 93:6	spam 188:22
48:20 124:12	198:12	104:11	104:13 105:20	spare 304:2
149:6	social 14:11	soloves 54:8	106:2 150:12	speak 10:5
sixteenplus	17:3 19:17	solution 315:14	152:19 159:17	20:10 25:11
125:1	27:2,8 37:12	316:10 358:21	172:19 175:20	117:10 124:16
size 58:9 101:19	197:16,22	solutions 115:18	180:10 192:14	151:22 249:5,7
152:14	199:7 229:10	285:19,20	210:14 218:5	295:22 303:16
skeptical 100:18	237:11 338:14	311:8,9 316:7	225:1 243:10	speaker 110:11
skew 73:12	338:22 340:3	341:21	246:2,17	speakers 383:4
skews 19:20	sociatial 227:21	solve 143:6	250:20 251:21	speaking 14:4
skill 305:10	societal 8:1 9:4	146:18	252:3 255:13	24:5 238:15
skin 35:8 37:6	19:13 20:2	somebody 47:7	259:20 260:17	265:15 284:7
			1	

360:2	sphere 56:7 85:5	198:9 222:18	374:3	steps 128:1
special 358:18	208:9,21	233:7 354:8	starts 47:11	131:8 132:6
359:1 376:7,13	210:21 211:1,7	358:19 359:22	283:7 379:12	172:12
376:18	spillway 351:2	367:10	stat 157:8	stewards 19:2
specialist 248:3	spoke 164:20,22	standards	state 26:4,5	stick 237:22
specialized	spoken 120:14	111:21 125:16	115:3 138:2,4	353:10
233:9	sponsors 311:11	125:16,20	157:20 187:16	stimulating
specific 36:15	spread 227:12	126:7 153:14	189:17 190:1	275:9
45:21 62:8	springing	163:2 196:19	220:19 302:20	stingrays 15:7
90:11 92:7,14	333:14	200:14,15	348:11 385:5	stockpiling
128:11 168:14	square 142:19	314:11	stated 262:21	111:19
172:15,16,17	squared 32:1	standarized	273:12	stolen 382:12
173:18 182:8	staff 7:14 110:6	232:14	statement	stood 278:8
214:4 220:15	116:21 193:20	standing 327:6	109:14 138:2	stool 345:10
234:15 244:15	242:9 248:7	start 10:6 23:22	276:10 277:6,7	stop 17:22 49:18
262:8 280:12	249:19 250:19	27:11 41:22	318:21,22	77:18,19 105:5
287:22 300:16	253:3,13,16	42:13 49:22	319:3 326:13	207:4,6 263:4
300:21 312:13	254:2,5,8,13	55:4 67:19	326:18 329:17	291:15
336:19 352:7,9	254:17,21	74:12 89:19	statements 99:9	storage 152:13
352:10 359:18	286:14,18	93:4 96:20,22	109:15 136:3	152:13
361:11	303:11 310:6,6	148:12 166:8	224:4	store 148:18,21
specifically	staffer 193:8,13	174:2 191:1	states 51:8	stored 19:8
42:11 68:4	193:16,17,19	192:11 207:5	127:22 193:8	103:13 221:3
98:22 153:19	303:18	250:10 255:7	228:4,6 343:22	storehouse
180:13 182:3	staffers 193:14	257:18 269:10	statistical 164:4	293:9
208:1 240:6,7	254:11 318:8	269:13,17	status 35:18	stories 16:17
247:3 280:11	staffs 245:9	273:13 274:17	37:12 229:22	story 69:2
281:19 311:6	stage 34:1,5,13	275:16,17,18	statutary 265:1	122:18 351:1
314:21 315:1	35:4 92:21	286:1,14,21	statute 301:19	straight 109:15
specification	174:9 372:20	287:6 306:5	statutorily	323:16
272:11 329:6	stages 57:18	312:2,20	281:7 282:11	strange 197:20
specificity 382:3	218:6	318:18 324:20	statutory 5:19	361:17
specifics 279:9	stake 59:10,12	338:13 343:21	122:10 204:17	strangest 59:13
specified 300:22	101:14	347:17 364:1	207:19 240:13	strategic 68:8
329:8	stakeholders	started 139:7	375:21	68:19 69:3
spectrum 205:6	308:21 361:9	198:14 210:19	stenographica	streamlined
speech 11:3	stand 153:11	242:7 250:11	385:7	132:17
355:7	194:13 233:19	304:6	step 46:18 76:16	street 148:19
speeches 9:15	286:15,16	starting 53:18	129:20 156:1	streets 311:20
spend 30:4	361:9 378:22	54:16 65:7	163:12 177:16	strength 376:12
233:16 239:7	standard 55:21	76:8 96:19	323:4 347:13	strengthen
275:5 313:9	70:17,18 74:16	100:22 159:18	354:19 369:18	225:20 281:7
spent 68:17	75:16 81:5	171:17 219:4	stepping 217:11	strengthening
333:21	82:17 167:4	274:15 283:8	354:13	131:9

stress 246:17	174:21 180:10	55:5	132:19 354:8	242:11 246:14
stretched 346:4	214:11 334:14	suffice 304:3	support 56:5	247:12,13
strict 6:22	348:15 367:19	sufficient 23:18	57:11 104:8	248:5 249:8
114:11 299:20	370:1	195:10 281:4	130:3 150:14	251:9,21
strides 234:8	subcommittee	300:5 341:10	187:3 237:5	252:11 254:21
strike 169:20	117:7	sufficiently	244:4 254:20	256:17 259:13
strikes 30:21	subject 130:8	253:14	312:6 350:4	265:11,18
77:12	137:13 141:18	suggest 132:20	354:20 366:13	267:3 278:13
striking 133:22	219:15 221:14	195:8 200:9	supported 56:1	289:8,12
134:5 337:8	222:15,17,21	236:16 276:14	126:20	303:20 311:13
strive 299:19	281:20 301:1	285:21 311:8		320:19 340:9
strive 299:19 strong 24:14	333:12	381:11	supporting 269:21	356:11 379:1
29:17 31:1	submissions			
	88:6	suggested 58:3 199:14 381:15	supports 227:15 231:4 234:14	surgical 317:5 365:12
111:3,7 115:16				
125:6 126:3	submit 6:14,16	suggesting	suppose 42:14	surgically 366:1
196:5 230:6	206:3 255:4	166:21 183:5	107:7 371:3,17	surpassing
235:1 268:5	384:4	288:5 350:21	supposed	111:2
280:18 299:3	submitted 7:10	350:22	272:18 285:18	surprised
369:19 378:22	110:5 318:22	suggestion	320:21 368:22	165:19 179:8
stronger 191:3	383:6	373:10	supposedly	241:11 242:20
195:10	subscriber	suggestions	302:22	290:3 301:13
strongly 29:9	169:7 175:14	165:5 305:6,15	supreme 14:12	surprises 38:2
128:10	subsequent	suggests 28:18	65:5,8 71:16	surprising
struck 53:7,13	131:16	57:10 108:10	76:8 150:14	301:14 330:8
88:22 89:17	substance 8:5	339:20	292:13 369:9	surrender 98:3
253:5 352:3	substantial	suit 298:6	374:1 375:1	surrendered
structure 21:20	36:12 68:13,22	299:17 327:7	sure 10:19 13:12	99:2 150:21
62:11 211:1	253:16	sum 15:1 105:20	18:21,22 41:20	surrendering
240:11	substantially	126:16 148:7	42:13 48:15,18	147:19
structured	302:1	summaries	50:10 61:19	surrenders
228:4 285:12	substitute	301:22	71:3 73:2 76:3	97:16 150:15
structures 258:3	348:12	summary	83:8 93:21	surround 348:5
struggle 197:5	substitutes	105:17	94:4,6 96:3	surrounding
struggling 87:11	288:19	summer 120:7	98:10 102:3	238:11
student 16:18	success 36:2	375:2	107:8 146:2,9	surveillance
145:19 229:11	237:9	super 134:11	157:3,6 159:2	5:12 16:16
students 157:9	successful	174:14	162:20 168:9	20:14,17,19
157:10	267:10,12	superb 287:15	173:7,15 175:3	21:12 23:12,14
studied 181:16	304:12	superbrief 76:5	175:17 185:3	23:21 24:11,18
studies 160:14	sudden 347:11	superviseory	203:13,15,18	32:7 44:14
study 145:1	suddenly 148:9	223:11	208:21 209:21	46:3 52:9,11
176:10	suffer 60:18	supervisory	210:22 211:2	67:13,14 72:16
stuff 120:17	suffered 292:5	219:11	211:19 232:20	81:1,6,18
152:20 154:9	suffering 54:6	supplement	237:18 240:16	84:14 85:5
	l ^v	l **	l	

87:15 127:22	62:11,20 77:3	47:4 52:5 61:9	218:3 219:22	target 35:8,20
128:11 129:7	85:10 112:13	63:20 64:16,17	225:20 238:2	62:6 141:11
130:5,8 148:10	135:8,17 138:4	68:16 72:6	268:13 272:19	174:7 228:16
148:17 153:15	140:20 155:21	75:21 104:5	276:17 326:14	256:14 262:20
167:2 168:13	156:11 157:18	109:2 156:1,13	326:16 330:15	targeted 44:13
180:14 186:6	158:19 170:5	156:16,20	341:12 348:14	67:13 82:6
		150.10,20		
212:11,21 229:19 249:1	174:11 183:9 187:14 188:21		382:13,16 talkative 139:4	181:3,4 230:1 271:13 301:7
295:19 249:1		163:17 170:4,4		303:6
	190:2,3 191:2 191:7,19 209:7	170:5,6 172:7 177:16 190:7	talked 46:5,8	
297:20 298:1	· · · · · · · · · · · · · · · · · · ·		53:21 54:3,14	targeting 279:2
300:13 301:11	209:8 220:15	225:3 227:1	60:22 63:22	targets 128:12
302:7,9 306:8	220:16,16	231:6 236:5	77:4 142:17	228:9,13 232:6
327:1 363:16	269:18 270:18	239:2 240:22	147:10 154:17	task 204:6
373:21 375:19	270:22 292:4	243:22 247:8	161:2 166:18	tax 13:14 69:9
surveilled 81:4	293:7 295:11	257:17 265:22	167:20 169:9	taxonomy 46:11
survellance	325:1,9 326:6	268:20 269:1	169:11,17	49:9,20
272:1	346:15,18	272:7 273:15	176:1 178:9,13	teach 311:18
survey 16:12	370:18	276:11 284:7	179:19,22	teaching 313:10
17:19	systemic 316:6	297:18 298:11	180:7 240:19	team 305:6
surveyed 367:1	324:19	303:19 316:8	245:10 256:7	322:1
survive 25:22	systems 37:15	317:17 321:8	268:9 282:1	tease 262:12
suspect 15:17	37:19,22 38:9	323:11 335:1	288:18 310:4	tech 3:3 110:14
329:19	95:19 112:1	347:13 362:10	319:1 326:13	143:10 250:21
suspended	134:17 135:3	362:11 379:13	348:8 373:1,17	technical 36:5
95:13	135:11,18	takeaways	376:7 381:20	36:17 40:22
suspicion 60:13	136:17 140:4	224:2	talking 21:22	41:5,9,10
suspicious 22:19	140:14 142:22	taken 4:19 69:6	29:3 68:15	62:15,17,19
sustainment	146:2,6,9	93:21 126:16	84:18 87:13	63:6,11 68:5
317:11 360:18	157:21 158:3	131:8 132:20	92:10 94:17	69:3 90:21
swath 324:15	158:17,20	169:5 170:18	100:1 101:16	91:5,16 92:1,7
swaths 282:19	170:22 173:1	172:9 205:10	119:13 137:14	95:19 104:8
sweeney 195:13	182:9 183:5	355:2 384:12	152:6 159:3	106:12,12
switch 348:16	185:1 201:16	takes 123:20	165:2 180:13	111:3 237:4,15
switches 39:16	214:15 230:9	171:21 176:21	185:9 203:16	249:14,20
synchronized	245:14,16	342:20 379:6	210:19 216:21	250:18 270:12
316:1	247:14 312:8	talismanically	233:16 271:16	279:13 305:10
synergy 39:6	312:17 315:17	288:11	272:5,17 275:5	309:19 351:22
synthesis 335:6	346:8,8 372:14	talk 9:10 54:20	329:5,6,7	360:4 376:15
347:1	372:18	64:15 67:11	340:2 341:18	378:7,21
synthesized	T	92:18 99:11	353:18 366:13	technically 91:1
323:10		108:1 119:6,17	talks 147:17	114:11 278:9
synthesizes	tagging 348:9	139:5 147:2	374:2	354:15,15
324:16	tailored 260:22	152:5 166:21	tangible 54:6	techniques
system 61:16	take 23:22 37:2	203:17 216:8	55:5 225:1	111:1 113:6

114:19 177:13	96:1,3 109:3	379:10,20	termed 54:1	198:6 369:15
232:7 321:13	109:13 110:17	380:8,11,13,17	terminology	testify 33:18
373:1	111:7,8 113:4	380:19,22	180:3	110:20 134:21
technological	113:8 117:5,8	381:1 383:14	terms 34:1 58:2	303:21
25:15,22 40:1	124:14 134:2	technologys	62:8 78:4	testimony 83:2
61:21 86:13	135:5,15	380:21	92:14 106:3	116:15 118:19
90:8 101:2	137:13 142:12	teenager 68:22	143:20 153:2	119:6,17 120:3
199:17 227:21	142:13,18	telephone 15:15	153:14 154:7,8	121:7 147:1
245:11,13	144:8 146:15	102:12,21	157:18 159:9	342:15
278:1 308:10	147:8,11,22	103:16 120:7	160:3 161:19	testing 231:13
340:6 350:1	148:3,4 149:3	163:17	164:17 168:16	252:13
353:21 379:2	149:16 155:1,6	television 22:12	184:15 185:22	tests 66:13
technologies	156:18 158:1	tell 98:17,18	218:18 223:1	text 12:19 136:7
40:6,14 78:19	162:2 165:3,11	123:2 166:20	245:15 248:20	136:7
105:13 106:4	167:8,11 170:2	180:6 184:18	252:9 261:3	thank 6:18 7:13
115:17 143:4	174:13,22	261:15,19	266:5 277:13	7:20 10:11
230:8 249:1,2	201:16 206:16	262:3 274:20	280:8 284:12	18:1 25:3,9,10
250:2 303:10	226:16,18	274:22 294:6	291:18,18	33:9 41:13,15
309:11 310:1	228:10,22	312:9 314:14	315:6 319:19	55:6 67:17
318:3 341:3	234:2,9,14,19	321:5 335:11	336:9 339:19	74:10,14 80:15
375:10 381:3	234:22 247:17	343:20 367:3,4	342:11 344:5	80:15 87:6
technologist	247:19 248:2,4	telling 15:13,18	348:9 356:17	88:14 96:8,8
112:11 115:14	248:11,14	99:13 293:21	358:5 359:2,16	102:1 108:19
116:2,4,16,18	250:14,16	303:17	361:4 373:8,14	109:5,8 116:9
134:22 160:11	251:5 252:10	tells 294:4	terrible 368:21	116:13,14
165:4 179:17	269:21 277:19	template 232:14	terribly 289:9	117:1,2,9,10
193:5 247:17	285:16 295:16	ten 94:9 167:10	territorial	124:9,10,15,15
248:8	296:8 304:12	181:22 205:20	220:19	134:9 141:15
technologists	305:14 306:2,7	288:6	terrorism 111:9	141:16 142:21
115:22 116:5,7	307:14 309:20	tenants 218:10	133:6 181:21	159:14,19
116:20 137:6,8	311:13,19	tend 66:20	182:4 183:4	165:13 171:9
144:22 145:6	314:4 315:11	77:16 267:19	217:2 296:3	171:10 172:5
146:4 154:19	315:18,19	tendency 61:1	365:16 371:15	196:7 202:20
162:10,12	320:10 325:2	tends 233:18	terrorismrelat	207:1,2,7,17
164:18 176:3,7	326:1 334:5	310:19	220:17	215:8 216:5,7
176:8,11,16,19	340:20 344:14	tenet 304:11	terrorist 15:18	216:7 225:8
237:11 239:6	344:15,19,19	tension 161:7	161:16 181:18	226:4,4 237:17
249:5 316:17	348:2,3 349:8	tenth 314:12,15	349:4	237:20 263:9
349:11 356:16	357:9 364:6,9	334:17	terrorists 36:1,4	271:4,5 278:2
357:3 383:10	364:9 373:14	terabytes	110:22 111:17	278:4,4 284:14
technology 3:2,5	373:17 374:18	349:21	113:1 166:2,15	285:9,10 287:8
3:19 6:2 15:7	374:20 375:5,5	term 95:13 97:2	tesla 26:6	293:16 295:12
33:14 39:21	377:10,21,22	152:4 222:2	105:11 106:2	295:13,19,21
40:9,10 61:12	378:9,15	260:2,3	test 147:15	296:1 303:7,8

303:15,16	143:9,10	338:2 348:11	153:19 154:9	199:2 224:22
313:3,8 318:7	147:14 151:17	349:1 350:3	156:15 164:7	251:3,12
322:13 326:7	151:21 152:8	353:5,6 354:2	171:1 172:7	267:17 283:16
328:16 347:16	151.21 152.8	354:5,21 356:7	173:6,10 175:6	283:16 287:1
347:19,19	161:10,11,12	357:13,20	177:6 183:9	310:7 311:18
363:1 374:5	161:17 162:1,2	358:13 360:15	184:19 185:19	312:1 315:13
381:6 383:1	166:12 167:6	363:4 366:16	192:22 193:3	326:19 328:9
384:14	169:6,22	366:21 368:11	193:13 197:12	332:19 334:17
	172:12,15,15	369:8 379:22		334:18 342:3
thankfully 193:21	172:12,13,13	theme 211:15	198:12,16 207:4 211:5	
				343:2,14,16
thanks 10:9	174:16 177:11	theoretical	220:2 221:18	344:6 357:6,7
33:17 67:21	177:12 178:4	285:14	253:9 258:4,5	357:8 363:15
107:19 108:22	179:18 181:6	theoretically	258:11 273:19	367:6 370:8
110:19 134:11	184:5,14 185:1	115:18	281:16 292:1	379:1
134:20 159:19	187:2 188:6,14	theory 38:15	292:21,22	theyve 24:10
169:8 284:20	189:6,14,19	76:7 78:5 89:1	293:1,2,2	33:20 106:10
284:22 303:14	195:22 198:4	89:3,11,12,18	308:3 317:1,2	149:14 177:5,7
313:2,6 383:3	199:6 200:17	97:5 250:12	324:13 328:4	289:8,10 367:2
thatd 160:12	206:9 210:14	268:13 368:5	329:1 332:12	371:6
thats 8:7 11:18	218:12 221:6	383:14	337:1 339:22	thin 79:12
12:2 13:20	222:9,15,20	thered 306:18	344:13 346:16	thing 10:14 14:3
14:5 15:10	223:6 224:19	theres 10:14	349:18 351:13	18:18 21:11
18:14 27:14,15	238:9 240:3	12:6 13:12,13	353:15 357:16	46:9 51:4 57:9
27:16,18 29:8	241:2 245:22	14:7 21:13	363:6 364:5	57:15 71:8,11
33:1 46:18	247:8 248:14	23:3 40:17	371:21	78:12 82:16
48:6,16 49:10	250:12 251:18	42:18 43:18	theyll 48:2	91:14 93:6
55:20 60:11,14	252:12 253:3	44:2 48:5,8,17	88:10	94:12 141:2
60:17 63:13	254:13 255:13	58:9 62:15	theyre 8:16 17:1	152:13 156:1,7
64:21 65:16	256:6 258:2,16	64:4,13 71:12	19:18 22:12	156:10,13
66:4,22 70:4	258:20 259:11	75:6 78:21	24:18 30:2,6	166:16 176:15
70:13 71:11	259:20 262:7	79:11 83:21	41:8 45:17,17	198:19 201:13
72:4,20 76:16	265:15 272:6	84:5,8 90:16	54:21 63:6,7	201:15 213:17
78:7 80:21	274:14 275:1,2	91:15,22 96:4	77:18 88:11	242:22 259:16
82:13,18 85:4	275:14 277:10	98:19,20 100:3	100:10,11	261:18 262:3
85:19 87:14	279:5 281:20	103:10,11,17	107:15 126:11	262:10 268:6
88:5,12 91:13	283:8 292:18	103:18 104:13	138:12,13,21	270:17 274:10
99:7,8 102:16	293:15 304:16	108:14 118:11	141:4 147:18	276:19 305:1
103:20 104:2	306:4 307:16	119:3 121:17	158:9 168:2	311:1 321:14
105:8 106:22	311:19 314:2	128:2 130:22	174:8,10,11	322:20 331:1
107:19 113:18	315:3 320:8	133:18 137:17	177:19 182:10	343:4 351:5
118:4,8 121:1	321:11,14	144:19 147:6	187:10 188:8	354:16 355:19
127:21 128:6	322:19,20	149:17 150:13	188:10 191:12	366:11
132:13 137:11	326:21 330:10	151:14,14	191:13,15,16	things 10:17,19
138:19 140:14	334:14 336:11	152:2,7 153:4	191:18 199:1,2	11:21 18:16,19
		I	l	l

19:5 21:8,13	334:9 335:3,13	83:13,14,17,19	196:15 197:11	292:1,15,18
25:1 28:12	336:4 339:4,16	83:20 85:2	197:15,20	293:11,13,18
29:15 30:14	349:14 351:16	86:8,9 87:9	198:8,15	294:1,11 295:1
32:17 42:22	356:17,18	88:18 89:10,16	199:14,19	295:2 305:16
43:1 44:11	357:5 359:5	90:1,20 92:8	200:2,13,17	306:6 309:16
46:5 49:17	361:15 379:3,8	93:5 94:12	205:1 206:15	311:10 313:14
50:3 52:3 53:6	381:19	96:4,6,9,13,20	207:20 208:6	314:2,4 315:3
59:13,15 68:9	thingsternet	97:1,11 98:15	210:15,18,20	317:14 319:6
68:9,13 71:6	139:2	99:11,22 100:6	211:14,17,21	317:14-317.0
84:4 86:9 93:7	think 7:21 10:3	101:1,5,13,19	212:3 213:3,5	321:11,21,22
94:10 95:18	10:21 11:8,8	101:1,5,15,15	213:14 220:1,6	327:9,21
104:13,14	14:10 17:16	102.16,20	222:9 225:13	328:19,21
105:3 115:1	19:9,19 20:10	103.3,21	225:14 227:20	329:3 330:4,17
105:3 115:1	21:18 25:19	104:12,13,16	233:20,22	331:5,9,14
			235:20,22	, ,
138:1,8,9,10	26:3,5,10,12	106:3,15		332:8,10 333:5
138:11 139:1,1	26:15,16 28:19	107:20 108:11	241:1,19	333:6 334:2,4
139:16 143:11	28:19 29:6	108:13 119:3	244:11,12,19	334:15,18,21
143:18 144:2	30:9,14 32:3	119:21 120:4	245:15,22	335:5,8 336:5
152:11,12,22	32:11,16,18	120:12,17,19	247:4 248:8,9	336:7,8,11,15
154:14 155:2	33:21,22 36:20	141:19 142:2,8	249:19 250:20	336:18,19
156:19 157:4,6	40:4 42:10,18	143:21 144:13	251:17 252:14	337:1,3 338:4
164:8 170:10	43:18 44:2,3,4	144:19 147:2,7	253:9,10 254:4	338:12,13,15
170:21 171:2,7	44:21 45:19	147:21 148:1,2	254:8,14	340:2,16,21
173:6,16 183:4	47:17,21 48:6	148:11 149:3,7	257:21 258:20	341:3,13 344:4
184:11 185:17	48:6,16 49:20	149:12,13	260:3,4,6,12	344:6 345:3,21
186:3,5 189:21	49:21 50:3	150:12 152:5	261:19,21	346:2,6 347:16
190:22 191:9	51:12 52:1,3	155:9 157:3,8	263:2,15	348:7,16 350:3
191:16 199:12	52:21,22 54:1	159:5 160:20	264:14,17	351:13 352:3
202:6,17	55:7 57:8,12	162:1,2,9	265:15 266:3	352:15 353:9
247:16 250:13	57:17,19 58:20	163:8,11,12,18	266:15 267:4,5	353:11 354:18
251:6,15	59:9,13 60:1,7	164:11 166:5	268:6,6 269:12	354:21 355:10
252:19 257:10	61:3,15,19	166:15,16	270:3,10 274:3	355:14,16
263:4 266:2	62:8 63:14,21	167:11,13,22	274:12,16	356:1,2,7,16
267:21 270:18	64:20 65:9	175:8,21 177:1	275:14 276:6	358:18,20
271:1 272:15	67:8,12,21	177:4,8,9	276:12 277:1,3	359:10,14,17
274:13 279:15	68:14 69:5	178:19,22	277:10,18	360:2 361:7
290:13 303:13	70:2 71:15,18	179:2,4,17,19	278:13 281:6,9	363:9 364:19
304:19 305:4	71:20 72:8,14	181:6 182:21	282:4,11,12,17	364:21 366:7,9
308:15,18	73:5,8,9,20	184:15 185:6	283:2,4,5,6,11	366:14 367:13
309:8,18,21	74:5,6,7,8,12	187:4,7 188:18	283:21 287:1,9	367:18 368:6,7
312:3,15 315:7	76:6,11 77:9	190:2,7,11,18	287:17 288:3	368:8 369:22
316:4 319:2	77:16 79:3,14	191:6,6,7,7,12	288:10 289:5	370:1,10
321:16 324:4	81:2,14 82:20	192:21 193:2,9	290:1,10,12	371:21 372:3,9
329:8 330:9	83:1,5,10,11	195:19 196:3,5	291:9,12,20	372:10,13,22
	l	ļ	· ·	

373:4,6 374:6	111:12	322:10 331:19	276:13 280:6	117:11 123:5
374:7,9,14,15	thought 32:14	346:20 360:16	286:7 287:9	124:16 125:22
374:7,5,14,15	51:4 54:3	360:19	297:10 310:20	134:7,21 151:7
374.20 373.0	55:14 83:4	threshold	310:22 313:9	181:22 205:1
376:9,13 377:5	84:10 87:18	276:15 322:22	319:14 320:15	205:16 216:21
377:7,11,16,17	90:7 164:15,16	thresholds	322:14,20	227:1 234:7
378:4,6,8,13	165:6 184:4	272:15,15,19	325:3 333:22	235:20 264:3
379:4,22 380:3	199:16 202:11	335:7	334:5 335:10	269:8 283:10
380:4,10,12	216:22 229:11	throw 44:18	336:3 342:20	285:3 295:22
381:9,14	256:1 261:10	63:20 275:17	344:17 346:3,7	303:16 305:13
382:18 383:7	273:2 276:22	throwaway	347:7,15,17,21	306:8 307:14
383:17	319:17 330:17	138:18	348:4 360:11	307:14 309:8
thinking 26:6,10	336:1 344:10	throwing	360:18 362:12	309:14 333:12
67:22 69:22	351:7,8 365:10	307:11	362:20 363:9	347:21 363:5
81:15 86:11	thoughtful	thunderbird	364:7 367:20	todays 5:15 7:6
101:14 143:17	383:8	26:2 40:5	368:1,18	7:17 10:13
143:18 210:19	thoughts 82:22	tight 159:12	370:11 373:19	33:22 35:10
229:6 250:4	87:8 153:4	tightly 267:9	379:6 380:17	123:21 340:20
292:16 295:8	183:7,18	time 6:16 9:13	381:7	383:22
308:9 312:2,15	187:21 189:7	9:20 15:2 17:6	timekeeper 6:22	toes 255:6
320:1 331:5	192:17,21	17:7 18:11	109:21	token 143:8
333:22 338:13	227:4 235:9	30:4 33:3	times 39:12	184:16 204:14
352:16 355:1	257:13 259:22	41:13 49:14,18	71:18 83:6	told 172:20
thinks 179:9,9	268:17 322:14	60:7 72:14	110:2 145:3	257:4
229:4	328:3,13,17	76:4 80:13	149:2 159:5	ton 178:1,1
thinner 45:1,1	366:10	84:2 87:21	190:16 241:10	tool 195:14
third 15:11	threads 316:6	113:15 116:21	257:4 273:12	289:2 291:1
20:12 34:3	threat 66:8 67:3	120:2 123:7,9	295:8 343:7	339:11 381:16
35:4 39:6 91:2	94:17 111:13	135:22 136:16	363:5	tools 117:15,16
111:13 119:17	114:15,19	137:18 138:17	timewise 75:22	126:12 133:5
130:3 150:16	115:12,19	139:6 141:18	tinker 196:2	143:16 173:8
151:5,17	199:20 217:13	144:9 146:1,11	tiny 120:10	175:1 177:13
168:16 229:4	218:15 226:14	148:13 164:1	tipping 256:19	190:12 233:9
283:14 285:17	227:20 228:1	164:21 168:1	260:2	234:9 314:7
291:22 302:3	threatens	175:9 190:8	tips 260:8,9	320:11 349:11
315:15 345:9	302:19	192:11 196:11	tire 303:17	382:9
thirdly 97:11	threats 115:8	205:21 206:2	title 109:11	top 81:7 82:2
thirdparty	133:4,7 227:11	214:22 215:1	219:9 247:3	150:4,5 162:7
13:21 72:4	228:6	218:4 220:21	today 5:17,20	181:11 272:18
97:13 197:13	three 34:1 91:10	223:22 226:8	6:19 9:3 25:12	296:13 349:5,6
353:15	111:6 123:14	227:1 231:19	26:3,7 33:18	topic 207:22
thirty 32:13	141:7 205:16	233:16 239:7	59:14 61:4	216:9 341:17
thomas 136:19	205:19 236:14	242:9 252:15	92:10 98:6	topics 16:15
thorough	263:13 256:11	255:7 275:5	105:12 111:6	383:11
		200 270.0	100.12 111.0	555.22

total 15:1 71:5 148:7 248:2 transborder transborder 153:8,15,15 350:7,8 345:9,10 356:5 transparent 89:5 208:18 214:6 298:19 trusted 242:10 trustworthy 155:17 234:11 truth 32:8 99:13 361:17 109:16 203:17 284:10 326:7 350:7,8 350:7,8 transcript 7:7 384:7 385:8 252:11 214:6 298:19 transperancy 212:1 truth 32:8 99:13 361:17 truned 82:8 truth 32:8 99:13 361:17 truned 82:8 249:10 26:22 100:622,19 40:60:17 truned 28:8 100:12 20:12 100:16:10:17 truned 28:8 100:17 truned 28:8 100:12:10:10:17 truned 28:8 100:12:10:10:17 trune 20:51 100:22:10:12:18 trune 30:11:18 trune 20:51 100:22:10:12:18 trune 20:51 100:22:10:22:19 trune 20:51 100:22:17:22:18 trune 20:51 100:22:17:22:18 trune 20:51 100:22:17:22:18 trune 20:51 100:22:17:22:18 trune 20:18:30:19 trune 20:18:30:19 trune 20:18:30:19 twitter 136:21 twitter 136:21 <th< th=""><th></th><th></th><th></th><th></th><th>1</th></th<>					1
148:7 totally 17:18 transborder 153:8,15,15 transparent 153:8,15,15 trustworthy 155:17 234:11 284:10 326:7 286:12 6:20 6:27,19 266:26,6 40:22:11 60:17 252:17 324:18 282:12 20:18 284:12 12:18 175:1178:16 284:10 326:12 284:10 326:17 284:10 326:17 284:10 326:17 284:10 326:17 284:10 326:17 284:10 326:17 284:10 326:17 284:10 326:17	total 15:1 71:5	248:2	345:9.10 356:5	trusted 242:10	109:16 203:17
totally 17:18 153:8,15,15 89:5 208:18 155:17 234:11 284:10 326:7 52:2 298:9 154:16 214:6 298:19 truth 32:8 99:13 tured 82:8 350:7,8 transcript 7:7 transcript 7:7 transperancy truth 32:8 99:13 turned 82:8 252:11 touched 170:10 90:17 212:1 try 45:10 56:22 20:18 321:2 turning 215:9 288:2 transition 60:20 88:20 107:22 trash 307:11 trace 415:18 155:15 173:14 tweeted 339:9,9 touching 139:9 88:20 107:22 transitions trace 415:18 175:1 178:16 tweeted 339:9,9 69:17 363:20 101:2 translate treat 21:18 175:1 178:16 tweeted 339:9,9 touted 183:11 103:11 340:18 treat 21:18 175:1 178:16 twice 70:9 track 71:16 171:14 tremendously trying 27:10 twitter 36:21 twitter 36:21 track 22:18 transparant 185:15,19 159:9 161:16 58:13 76:17 twitter 36:21 trade 216:12,15 34:1 transparant 186:17 187:2			<i>'</i>		
52:2 298:9 154:16 214:6 298:19 truth 32:8 99:13 turned 82:8 350:7,8 transcript 7:7 384:7 385:8 122:11 transperancy 212:1 try 45:10 56:22 205:18 321:2 205:18 321:2 205:18 321:2 truning 215:9 truning 21:9 truning 21:9 truning 21:2 48:13 361:31 truning 21:2 249:19 262:12 26:212 26:212 26:212 26:212 26:212 2			_	•	
350:7,8 touch 127:8 transcript 7:7 384:7 385:8 transferred yestem touched 170:10 transferred yestem touched 170:10 348:7 385:8 transferred yestem touched 170:10 361:17 yestem designed yestem touching 139:9 yestem touching 139:9 transiting 255:1 transition 60:20 transitions 69:17 363:20 transition 60:20 translate 88:19 translations 101:2 translating 101:2 translating 101:2 translating 101:2 translating 173:11 translating 101:2 translating 173:11 translating 142:19 translation 33:16 tremendous 142:19 translation 33:16 tremendous 142:19 translation 33:17 trend 142:4 translatior 185:15,19 translatior 185:15,19 <t< th=""><th></th><th>, ,</th><th></th><th></th><th></th></t<>		, ,			
touch 127:8 384:7 385:8 transferred with fouched 170:10 212:1 transportation fouched 170:10 transferred stransportation fouched 170:10 40:120 62:7,19 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:9 turning 215:14 turning 215:9 turning 215:9 turning 215:9 turning 215:14 turning 215:					122:21 202:12
252:11	II '				
touched 170:10 288:2 90:17 transition 60:20 139:11 59:20,22 60:5 trash 307:11 68:2,6,6 105:21 160:17 165:15 173:14 turns 250:14 252:7322:8 turns 250:14 252:7322:8 touchstone 69:17 363:20 transitions 101:2 treated 150:8 231:7 249:19 262:12 231:7 337:17 tweeted 339:9,9 touted 183:11 183:17 translate 88:19 translating treaties 51:8,13 142:19 323:16 340:21 348:13 361:13 twite 70:9 twiez 70:9 town 296:5 171:14 translator 378:14 tremendous tremendously trend 142:4 trying 27:10 828:4 twitter 136:21 68:14 72:21 twitter 136:21 twitter 337:16 track 71:16 280:6 378:14 trend 142:4 trend 142:4 82:14 84:22 82:14 84:22 twitter 36:21 4 two 91:2 30:15 trade 216:12,15 314:1 139:9 208:13 trends 144:16 171:5 175:5 191:18 196:1 91:18 196:1 116:20 119:13 tradecraft 187:17 transparency 23:2,10 28:5 tried 36:1 85:12 137:19 239:8 242:3 137:19 122:4 124:21 199:10 211:5 199:10 21:15 199:10 21:15 traffic 39:10 85:15 86:2 127:618 true 17:9 48:6 67:9 255:13 289:3,5 274:4 276:1 275:23 289:3,5 274:4 276:1 276:11 286:1 199:10 211:15 train 174:5 165:10 204:16 17:15 172:12<			transportation	•	turning 215:9
288:2 touching 139:9 139:11 transition 60:20 139:11 transition 60:20 travel 9:5 tract 121:18 transition 175:1 173:14 travel 9:5 tract 121:18 transition 175:1 178:16 travel 9:5 tract 121:18 transition 60:20 tract 121:18 transition 175:1 178:16 tweeted 339:9,9 twentyfirst 337:17 tweeted 339:9,9 twentyfirst 337:17 toughest 327:2 touted 183:11 101:2 231:7 269:13 316:9 treeted 150:8 249:19 262:12 337:17 twice 70:9 treeted 150:8 337:17 twice 70:9 treeted 150:8 249:19 262:12 337:16 tweeted 339:9,9 tweeted 339:9,9 treeted 150:8 tract 249:19 262:12 337:17 twice 70:9 treeted 150:8 337:17 twice 70:9 treeted 150:8 249:19 262:12 337:16 tweeted 330:16 340:21 treeted 150:8 142:19 27:10 treeted 150:8 249:19 262:12 378:15 twiter 70:9 treeted 150:8 337:16 twiter 337:16 twiter 337:16 twiter 337:16 twiter 136:21 78:15 78:1	touched 170:10	90:17	_	,	O
touching 139:9 transition 60:20 travel 9:5 165:15 173:14 tweeted 339:9,9 139:11 88:20 107:22 treat 121:18 175:1 178:16 337:17 touchstone 69:17 363:20 101:2 231:7 269:13 316:9 twice 70:9 touted 183:11 305:11 340:18 treaties 51:8,13 323:16 340:21 twist 337:16 touted 183:17 translating 142:19 378:15 twist 337:16 town 296:5 171:14 tremendousy tring 17:10 twitter 368:4 track 71:16 288:4 68:14 72:21 two 9:12 30:15 280:6 378:14 trend 142:4 82:14 84:22 34:14 43:22 track 22:18 139:9 trends 144:16 171:5 175:5 91:9 105:9,17 trade 216:12,15 transparent 186:7 187:22 19:18 196:1 10:11 14:17 187:17 23:2,10 28:5 137:19 243:6 247:10 140:7 186:2 traditional 31:13,41,71,8 trigger 60:13 25:14 256:2 187:22 194:6 274:8 32:9 84:5 85:8 troup 18:3	288:2	transiting 255:1	trash 307:11	105:21 160:17	252:7 322:8
touchstone transitions treated 150:8 249:19 262:12 337:17 twice 70:9 69:17 363:20 toughest 327:2 translate 88:19 treaties 51:8,13 323:16 340:21 twice 70:9 touted 183:11 305:11 340:18 tremendous 348:13 361:13 twitter 136:21 183:17 translating 142:19 378:15 twitter 136:21 town 296:5 171:14 tremendously trying 27:10 twitter 368:4 track 71:16 translator 288:4 68:14 72:21 two 9:12 30:15 280:6 378:14 trend 142:4 185:15,19 159:9 161:16 58:13 76:17 tracks 222:18 transparent 186:7 187:22 191:18 196:1 110:1 114:17 tradecaft transparency tried 36:1 85:12 199:17 213:7,9 116:20 119:13 traditional 31:13,14,17,18 232;0 28:5 137:19 243:6 247:10 140:7 186:2 274:8 32:9 84:5 85:8 triger 60:13 25:14 256:2 187:22 194:6 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:2	touching 139:9		travel 9:5	165:15 173:14	tweeted 339:9,9
69:17 363:20 101:2 231:7 269:13 316:9 twice 70:9 toughest 327:2 translate 88:19 treaties 51:8,13 323:16 340:21 twite 70:9 touted 183:11 translating 142:19 348:13 36:13 twitter 136:21 town 296:5 171:14 tremendously trying 27:10 twitter 368:4 track 71:16 280:6 378:14 trend 142:4 82:14 84:22 twitters 368:4 tracking 306:19 transmitted 185:15,19 159:9 161:16 58:13 76:17 trade 216:12,15 transparant 186:7 187:22 199:17 213:7,9 116:20 119:13 tradecraft transparency tried 36:1 85:12 137:19 243:6 247:10 140:7 186:2 traditional 31:13,14,17,18 204:9 232:4 31:20,22 32:4 trigger 60:13 251:14 256:2 179:10 211:15 trail 276:18 89:4 125:22 179:7 30:13 315:17 316:19 277:3 3289:3,5 274:4 276:1 trail 276:18 89:4 125:22 199:7 130:13 315:17 316:19 279:11 286:5 trail 276:18 29:17 212:2	139:11	88:20 107:22	treat 121:18	175:1 178:16	twentyfirst
toughest 327:2 touted 183:11 translate 88:19 305:11 340:18 translating treaties 51:8,13 142:19 378:15 367:22 trying 27:10 378:15 367:22 trying 27:10 trying	touchstone	transitions	treated 150:8	249:19 262:12	337:17
touted 183:17 305:11 340:18 tremendous 348:13 361:13 twitter 136:21 town 296:5 171:14 tremendously trying 27:10 twitters 368:4 track 71:16 translator 288:4 68:14 72:21 twitters 368:4 280:6 378:14 trend 142:4 82:14 84:22 34:14 3:22 tracking 306:19 transmitted 185:15,19 159:9 161:16 58:13 76:17 tracks 222:18 transparant 186:7 187:22 191:18 196:1 110:1 114:17 314:1 208:13 tribal 220:19 199:17 213:7,9 116:20 119:13 tradecraft transparency tried 36:1 85:12 239:8 242:3 122:4 124:21 187:17 23:2,10 28:5 137:19 243:6 247:10 140:7 186:2 traditional 31:13,14,17,18 trigger 60:13 25:114 256:2 187:29 19:10 21:15 274:8 32:9 84:5 85:8 troops 227:15 264:10,12 217:22 250:17 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 278:8,11 trails 93:3 94:4 128:22,22 <th>69:17 363:20</th> <th>101:2</th> <th>231:7</th> <th>269:13 316:9</th> <th>twice 70:9</th>	69:17 363:20	101:2	231:7	269:13 316:9	twice 70:9
183:17 translating 142:19 378:15 367:22 town 296:5 171:14 tremendously 288:4 68:14 72:21 twitters 368:4 track 71:16 378:14 trend 142:4 82:14 84:22 34:14 43:22 tracking 306:19 tracks 222:18 transmitted 185:15,19 159:9 161:16 58:13 76:17 trade 216:12,15 transparant 186:7 187:22 191:18 196:1 110:1 114:17 314:1 208:13 tribal 220:19 199:17 213:7,9 116:20 119:13 tradecraft transparency tried 36:1 85:12 239:8 242:3 122:4 124:21 187:17 23:2,10 28:5 137:19 243:6 247:10 140:7 186:2 187:22 194:6 traditional 31:13,14,17,18 trigger 60:13 251:14 256:2 187:22 194:6 274:8 32:9 84:5 85:8 troops 227:15 264:10,12 217:22 250:17 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 278:8,11 train 174:5 163:8 164:17 truly 40:20,21 334:3 342:12 334:3 342:12 334:3 342:12 <th>toughest 327:2</th> <th>translate 88:19</th> <th>treaties 51:8,13</th> <th>323:16 340:21</th> <th>twist 337:16</th>	toughest 327:2	translate 88:19	treaties 51:8,13	323:16 340:21	twist 337:16
town 296:5 171:14 tremendously trying 27:10 twitters 368:4 track 71:16 280:6 378:14 trend 142:4 82:14 84:22 34:14 43:22 tracking 306:19 transmitted 185:15,19 159:9 161:16 58:13 76:17 tracks 222:18 139:9 trends 144:16 171:5 175:5 91:9 105:9,17 trade 216:12,15 transparant 186:7 187:22 191:18 196:1 110:1 114:17 314:1 208:13 tribal 220:19 199:17 213:7,9 91:9 105:9,17 tradecraft transparency tried 36:1 85:12 239:8 242:3 122:4 124:21 187:17 23:2,10 28:5 137:19 243:6 247:10 140:7 186:2 traditional 31:13,14,17,18 trigger 60:13 trillion 141:7 263:15,21 199:10 211:15 24:8 32:9 84:5 85:8 troops 227:15 264:10,12 217:22 250:17 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 275:3 289:3,5 274:4 276:1 train 174:5 163:8 164:17 151:2 1 336:19 320:2,3 328:22 336	touted 183:11	305:11 340:18	tremendous	348:13 361:13	twitter 136:21
track 71:16 translator 288:4 68:14 72:21 two 9:12 30:15 280:6 378:14 trend 142:4 82:14 84:22 34:14 43:22 tracking 306:19 transmitted 139:9 trends 144:16 171:5 175:5 91:9 105:9,17 trade 216:12,15 transparant 186:7 187:22 191:18 196:1 110:1 114:17 314:1 208:13 tribal 220:19 199:17 213:7,9 116:20 119:13 tradecraft transparency tried 36:1 85:12 239:8 242:3 122:4 124:21 187:17 23:2,10 28:5 137:19 243:6 247:10 140:7 186:2 traditional 31:13,14,17,18 trigger 60:13 251:14 256:2 187:22 194:6 204:9 232:4 31:20,22 32:4 trillion 141:7 263:15,21 199:10 211:15 traffic 39:10 85:15 86:2 troops 227:15 264:10,12 217:22 250:17 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 278:8,11 trail 276:18 89:4 125:22 99:7 130:13 315:17 316:19 279:11 286:5 train 174:5	183:17	translating	142:19	378:15	367:22
280:6 378:14 trend 142:4 82:14 84:22 34:14 43:22 tracking 306:19 transmitted 185:15,19 159:9 161:16 58:13 76:17 tracks 222:18 transparant 186:7 187:22 191:18 196:1 110:1 114:17 314:1 208:13 tribal 220:19 199:17 213:7,9 116:20 119:13 tradecraft transparency tried 36:1 85:12 239:8 242:3 122:4 124:21 187:17 23:2,10 28:5 137:19 243:6 247:10 140:7 186:2 traditional 31:13,14,17,18 trigger 60:13 251:14 256:2 187:22 194:6 204:9 232:4 31:20,22 32:4 trillion 141:7 263:15,21 199:10 211:15 274:8 32:9 84:5 85:8 troops 227:15 264:10,12 217:22 250:17 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 278:8,11 trail 39:2 130:11 152:16 151:21 336:19 320:2,3 328:22 306:1 309:5 train 174:5 163:8 164:17 truly 40:20,21 343:3 434:15 334:6 326:10 172:14 173:20 207:12	town 296:5	171:14		· ·	
tracking 306:19 transmitted 185:15,19 159:9 161:16 58:13 76:17 tracks 222:18 139:9 trends 144:16 171:5 175:5 91:9 105:9,17 trade 216:12,15 transparant 186:7 187:22 191:18 196:1 110:1 114:17 314:1 transparency tried 36:1 85:12 239:8 242:3 122:4 124:21 187:17 23:2,10 28:5 137:19 243:6 247:10 140:7 186:2 traditional 31:13,14,17,18 trigger 60:13 251:14 256:2 187:22 194:6 204:9 232:4 31:20,22 32:4 trillion 141:7 263:15,21 199:10 211:15 274:8 32:9 84:5 85:8 troops 227:15 264:10,12 217:22 250:17 trail 276:18 89:4 125:22 p9:7 130:13 315:17 316:19 279:11 286:5 319:2 130:11 152:16 151:21 336:19 320:2,3 328:22 306:1 309:5 train 174:5 163:8 164:17 truly 40:20,21 343:13 342:12 314:6 326:10 212:15 175:172:12 210:17 211:9 trust 24:10 58:1 378:18 36:18 36:1 172:14 173:20					
tracks 222:18 139:9 trends 144:16 171:5 175:5 91:9 105:9,17 trade 216:12,15 transparant 186:7 187:22 191:18 196:1 110:1 114:17 314:1 208:13 tribal 220:19 199:17 213:7,9 116:20 119:13 tradecraft transparency tried 36:1 85:12 239:8 242:3 122:4 124:21 187:17 23:2,10 28:5 137:19 243:6 247:10 140:7 186:2 traditional 31:13,14,17,18 trigger 60:13 251:14 256:2 187:22 194:6 204:9 232:4 31:20,22 32:4 trillion 141:7 263:15,21 199:10 211:15 274:8 32:9 84:5 85:8 troops 227:15 264:10,12 217:22 250:17 traffic 39:10 85:15 86:2 troubling 80:12 275:3 289:3,5 274:4 276:1 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 278:8,11 train 174:5 163:8 164:17 151:21 336:19 320:2,3 328:22 306:1 309:5 training 94:22 207:12 209:6 67:9 345:7 376:11 345:13 36:2 172:14 173:20 <t< th=""><th>280:6</th><th>378:14</th><th>trend 142:4</th><th>82:14 84:22</th><th>34:14 43:22</th></t<>	280:6	378:14	trend 142:4	82:14 84:22	34:14 43:22
trade 216:12,15 transparant 186:7 187:22 191:18 196:1 110:1 114:17 314:1 208:13 tribal 220:19 199:17 213:7,9 116:20 119:13 tradecraft transparency tried 36:1 85:12 239:8 242:3 122:4 124:21 187:17 23:2,10 28:5 137:19 243:6 247:10 140:7 186:2 traditional 31:13,14,17,18 trigger 60:13 251:14 256:2 187:22 194:6 204:9 232:4 31:20,22 32:4 trillion 141:7 263:15,21 199:10 211:15 274:8 32:9 84:5 85:8 troops 227:15 264:10,12 217:22 250:17 traffic 39:10 85:15 86:2 troubling 80:12 275:3 289:3,5 274:4 276:1 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 278:8,11 trails 93:3 94:4 128:22,22 99:7 130:13 315:17 316:19 279:11 286:5 319:2 130:11 152:16 151:21 336:19 320:2,3 328:22 306:1 309:5 train 174:5 163:8 164:17 truly 40:20,21 334:3 342:12 314:6 326:10 212:5 172:12	$\overline{\mathcal{C}}$	transmitted	/	159:9 161:16	58:13 76:17
314:1 208:13 tribal 220:19 199:17 213:7,9 116:20 119:13 tradecraft 187:17 23:2,10 28:5 137:19 243:6 247:10 140:7 186:2 traditional 204:9 232:4 31:13,14,17,18 trigger 60:13 251:14 256:2 187:22 194:6 204:9 232:4 31:20,22 32:4 trillion 141:7 263:15,21 199:10 211:15 274:8 32:9 84:5 85:8 troops 227:15 264:10,12 217:22 250:17 traffic 39:10 85:15 86:2 troubling 80:12 275:3 289:3,5 274:4 276:1 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 278:8,11 trails 93:3 94:4 128:22,22 99:7 130:13 315:17 316:19 279:11 286:5 319:2 130:11 152:16 151:21 336:19 320:2,3 328:22 306:1 309:5 train 174:5 163:8 164:17 truly 40:20,21 334:3 342:12 314:6 326:10 214:15 165:10 204:16 67:9 345:7 376:11 345:13 361:2 172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 371:4 372:19 215:3 244:1	tracks 222:18	139:9	trends 144:16	171:5 175:5	,
tradecraft transparency tried 36:1 85:12 239:8 242:3 122:4 124:21 187:17 23:2,10 28:5 137:19 243:6 247:10 140:7 186:2 traditional 31:13,14,17,18 trigger 60:13 251:14 256:2 187:22 194:6 204:9 232:4 31:20,22 32:4 trillion 141:7 263:15,21 199:10 211:15 274:8 32:9 84:5 85:8 troops 227:15 264:10,12 217:22 250:17 traffic 39:10 85:15 86:2 troubling 80:12 275:3 289:3,5 274:4 276:1 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 278:8,11 trails 93:3 94:4 128:22,22 99:7 130:13 315:17 316:19 279:11 286:5 319:2 130:11 152:16 151:21 336:19 320:2,3 328:22 306:1 309:5 train 174:5 163:8 164:17 truly 40:20,21 334:3 342:12 314:6 326:10 214:15 165:10 204:16 67:9 345:7 376:11 345:13 36:12 172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 378:18 361:18 362:1	II '	_			
187:17 23:2,10 28:5 137:19 243:6 247:10 140:7 186:2 traditional 31:13,14,17,18 251:14 256:2 187:22 194:6 204:9 232:4 31:20,22 32:4 trillion 141:7 263:15,21 199:10 211:15 274:8 32:9 84:5 85:8 troops 227:15 264:10,12 217:22 250:17 traffic 39:10 85:15 86:2 troubling 80:12 275:3 289:3,5 274:4 276:1 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 278:8,11 trails 93:3 94:4 128:22,22 99:7 130:13 315:17 316:19 279:11 286:5 319:2 130:11 152:16 151:21 336:19 320:2,3 328:22 306:1 309:5 train 174:5 163:8 164:17 truly 40:20,21 334:3 342:12 314:6 326:10 214:15 165:10 204:16 trump 66:14 343:14 344:5 334:9 336:4 training 94:22 207:12 209:6 67:9 345:7 376:11 345:13 361:2 172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 371:4 372:19 215:3 244:15 214:13 215:2		208:13			
traditional 31:13,14,17,18 trigger 60:13 251:14 256:2 187:22 194:6 204:9 232:4 31:20,22 32:4 trillion 141:7 263:15,21 199:10 211:15 274:8 32:9 84:5 85:8 troops 227:15 264:10,12 217:22 250:17 traffic 39:10 85:15 86:2 troubling 80:12 275:3 289:3,5 274:4 276:1 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 278:8,11 trails 93:3 94:4 128:22,22 99:7 130:13 315:17 316:19 279:11 286:5 319:2 130:11 152:16 151:21 336:19 320:2,3 328:22 306:1 309:5 train 174:5 163:8 164:17 truly 40:20,21 334:3 342:12 314:6 326:10 214:15 165:10 204:16 trump 66:14 343:14 344:5 334:9 336:4 training 94:22 207:12 209:6 67:9 345:7 376:11 345:13 36:12 172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 371:4 372:19 215:3 244:15 214:13 215:2 115:12 125:7 83:13 tssi 193:8,15,16 type 31:16					
204:9 232:4 31:20,22 32:4 trillion 141:7 263:15,21 199:10 211:15 274:8 32:9 84:5 85:8 troops 227:15 264:10,12 217:22 250:17 traffic 39:10 85:15 86:2 troubling 80:12 275:3 289:3,5 274:4 276:1 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 278:8,11 trails 93:3 94:4 128:22,22 99:7 130:13 315:17 316:19 279:11 286:5 319:2 130:11 152:16 151:21 336:19 320:2,3 328:22 306:1 309:5 train 174:5 163:8 164:17 truly 40:20,21 343:13 342:12 314:6 326:10 214:15 165:10 204:16 67:9 345:7 376:11 345:13 361:2 172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 371:4 372:19 215:3 244:15 214:13 215:2 115:12 125:7 83:13 380:9 378:4 220:7 271:19 125:11,12,14 tssi 193:8,15,16 type 31:16 transaction 272:3 274:8 127:12,17,20 193:22 201:6 129:21 168:11 331:7 280:14 298		<i>'</i>			
274:8 32:9 84:5 85:8 troops 227:15 264:10,12 217:22 250:17 traffic 39:10 85:15 86:2 troubling 80:12 275:3 289:3,5 274:4 276:1 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 278:8,11 trails 93:3 94:4 128:22,22 99:7 130:13 315:17 316:19 279:11 286:5 319:2 130:11 152:16 151:21 336:19 320:2,3 328:22 306:1 309:5 train 174:5 163:8 164:17 truly 40:20,21 343:13 342:12 314:6 326:10 214:15 165:10 204:16 trump 66:14 343:14 344:5 334:9 336:4 training 94:22 207:12 209:6 67:9 345:7 376:11 345:13 361:2 172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 371:4 372:19 215:3 244:15 214:13 215:2 115:12 125:7 83:13 tspe 31:16 transaction 272:3 274:8 127:12,17,20 193:22 201:6 129:21 168:11 331:7 280:14 298:16 128:5 135:16 tuning 182:18 251:10 262:17 transactional		, , ,	00		
traffic 39:10 85:15 86:2 troubling 80:12 275:3 289:3,5 274:4 276:1 trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 278:8,11 trails 93:3 94:4 128:22,22 99:7 130:13 315:17 316:19 279:11 286:5 319:2 130:11 152:16 151:21 336:19 320:2,3 328:22 306:1 309:5 train 174:5 163:8 164:17 truly 40:20,21 334:3 342:12 314:6 326:10 214:15 165:10 204:16 trump 66:14 343:14 344:5 334:9 336:4 training 94:22 207:12 209:6 67:9 345:7 376:11 345:13 361:2 172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 371:4 372:19 215:3 244:15 214:13 215:2 115:12 125:7 83:13 380:9 type 31:16 transaction 272:3 274:8 127:12,17,20 193:22 201:6 129:21 168:11 transactional 301:10 302:1,4 247:12 249:11 turn 7:18 28:18 251:10 262:17 152:1 320:4 303:7 327:16 265:13 266:13 34:21 36:22 302:10 331:16		· ·		<i>'</i>	
trail 276:18 89:4 125:22 true 17:9 48:6 291:15 313:21 278:8,11 319:2 130:11 152:16 151:21 336:19 320:2,3 328:22 306:1 309:5 train 174:5 163:8 164:17 truly 40:20,21 334:3 342:12 314:6 326:10 214:15 165:10 204:16 trump 66:14 343:14 344:5 334:9 336:4 training 94:22 207:12 209:6 67:9 345:7 376:11 345:13 361:2 172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 371:4 372:19 215:3 244:15 214:13 215:2 115:12 125:7 83:13 380:9 378:4 220:7 271:19 125:11,12,14 tssi 193:8,15,16 type 31:16 transaction 272:3 274:8 127:12,17,20 193:22 201:6 129:21 168:11 331:7 280:14 298:16 128:5 135:16 tuning 182:18 251:10 262:17 transactional 301:10 302:1,4 247:12 249:11 turn 7:18 28:18 251:10 262:17 152:1 320:4 303:7 327:16 265:13 266:13 34:21 36:22 302:10 331:16	1		_	· · · · · · · · · · · · · · · · · · ·	
trails 93:3 94:4 128:22,22 99:7 130:13 315:17 316:19 279:11 286:5 319:2 130:11 152:16 151:21 336:19 320:2,3 328:22 306:1 309:5 train 174:5 163:8 164:17 truly 40:20,21 334:3 342:12 314:6 326:10 214:15 165:10 204:16 trump 66:14 343:14 344:5 334:9 336:4 training 94:22 207:12 209:6 67:9 345:7 376:11 345:13 361:2 172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 371:4 372:19 215:3 244:15 214:13 215:2 115:12 125:7 83:13 380:9 378:4 220:7 271:19 125:11,12,14 tssi 193:8,15,16 type 31:16 transaction 272:3 274:8 127:12,17,20 193:22 201:6 129:21 168:11 331:7 280:14 298:16 128:5 135:16 tuning 182:18 251:10 262:17 transactional 301:10 302:1,4 247:12 249:11 turn 7:18 28:18 251:10 262:17 152:1 320:4 303:7 327:16 265:13 266:13 34:21 36:22 302:10 331:16			$\overline{\mathcal{C}}$		
319:2 130:11 152:16 151:21 336:19 320:2,3 328:22 306:1 309:5 train 174:5 163:8 164:17 truly 40:20,21 334:3 342:12 314:6 326:10 214:15 165:10 204:16 trump 66:14 343:14 344:5 334:9 336:4 training 94:22 207:12 209:6 67:9 345:7 376:11 345:13 361:2 127:5 172:12 210:17 211:9 trust 24:10 58:1 378:18 361:18 362:1 172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 371:4 372:19 215:3 244:15 214:13 215:2 115:12 125:7 83:13 380:9 125:11,12,14 transaction 272:3 274:8 127:12,17,20 193:22 201:6 129:21 168:11 331:7 280:14 298:16 128:5 135:16 tuning 182:18 239:5 248:18 transactional 301:10 302:1,4 247:12 249:11 34:21 36:22 302:10 331:16					,
train 174:5 163:8 164:17 truly 40:20,21 334:3 342:12 314:6 326:10 214:15 165:10 204:16 trump 66:14 343:14 344:5 334:9 336:4 training 94:22 207:12 209:6 67:9 345:7 376:11 345:13 361:2 127:5 172:12 210:17 211:9 trust 24:10 58:1 378:18 361:18 362:1 172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 371:4 372:19 215:3 244:15 214:13 215:2 115:12 125:7 83:13 380:9 378:4 220:7 271:19 125:11,12,14 tssi 193:8,15,16 type 31:16 transaction 272:3 274:8 127:12,17,20 193:22 201:6 129:21 168:11 331:7 280:14 298:16 128:5 135:16 tuning 182:18 239:5 248:18 transactional 301:10 302:1,4 247:12 249:11 turn 7:18 28:18 251:10 262:17 152:1 320:4 303:7 327:16 265:13 266:13 34:21 36:22 302:10 331:16		,			
214:15 165:10 204:16 trump 66:14 343:14 344:5 334:9 336:4 training 94:22 207:12 209:6 67:9 345:7 376:11 345:13 361:2 127:5 172:12 210:17 211:9 trust 24:10 58:1 378:18 361:18 362:1 172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 371:4 372:19 215:3 244:15 214:13 215:2 115:12 125:7 83:13 380:9 378:4 220:7 271:19 125:11,12,14 tssi 193:8,15,16 type 31:16 transaction 272:3 274:8 127:12,17,20 193:22 201:6 129:21 168:11 331:7 280:14 298:16 128:5 135:16 tuning 182:18 239:5 248:18 transactional 301:10 302:1,4 247:12 249:11 turn 7:18 28:18 251:10 262:17 152:1 320:4 303:7 327:16 265:13 266:13 34:21 36:22 302:10 331:16	II			· /	
training 94:22 207:12 209:6 67:9 345:7 376:11 345:13 361:2 127:5 172:12 210:17 211:9 trust 24:10 58:1 378:18 361:18 362:1 172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 371:4 372:19 215:3 244:15 214:13 215:2 115:12 125:7 83:13 380:9 378:4 220:7 271:19 125:11,12,14 tssi 193:8,15,16 type 31:16 transaction 272:3 274:8 127:12,17,20 193:22 201:6 129:21 168:11 331:7 280:14 298:16 128:5 135:16 tuning 182:18 239:5 248:18 transactional 301:10 302:1,4 247:12 249:11 turn 7:18 28:18 251:10 262:17 152:1 320:4 303:7 327:16 265:13 266:13 34:21 36:22 302:10 331:16					
127:5 172:12 210:17 211:9 trust 24:10 58:1 378:18 361:18 362:1 172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 371:4 372:19 215:3 244:15 214:13 215:2 115:12 125:7 83:13 380:9 378:4 220:7 271:19 125:11,12,14 tssi 193:8,15,16 type 31:16 transaction 272:3 274:8 127:12,17,20 193:22 201:6 129:21 168:11 331:7 280:14 298:16 128:5 135:16 tuning 182:18 239:5 248:18 transactional 301:10 302:1,4 247:12 249:11 turn 7:18 28:18 251:10 262:17 152:1 320:4 303:7 327:16 265:13 266:13 34:21 36:22 302:10 331:16			_		
172:14 173:20 213:15 214:2 58:4 85:15 tsa 58:14 59:16 371:4 372:19 215:3 244:15 214:13 215:2 115:12 125:7 83:13 380:9 378:4 220:7 271:19 125:11,12,14 tssi 193:8,15,16 type 31:16 transaction 272:3 274:8 127:12,17,20 193:22 201:6 129:21 168:11 331:7 280:14 298:16 128:5 135:16 tuning 182:18 239:5 248:18 transactional 301:10 302:1,4 247:12 249:11 turn 7:18 28:18 251:10 262:17 152:1 320:4 303:7 327:16 265:13 266:13 34:21 36:22 302:10 331:16					
215:3 244:15 214:13 215:2 115:12 125:7 83:13 380:9 378:4 220:7 271:19 125:11,12,14 tssi 193:8,15,16 type 31:16 transaction 272:3 274:8 127:12,17,20 193:22 201:6 129:21 168:11 331:7 280:14 298:16 128:5 135:16 tuning 182:18 239:5 248:18 transactional 301:10 302:1,4 247:12 249:11 turn 7:18 28:18 251:10 262:17 152:1 320:4 303:7 327:16 265:13 266:13 34:21 36:22 302:10 331:16					
378:4 220:7 271:19 125:11,12,14 tssi 193:8,15,16 type 31:16 transaction 272:3 274:8 127:12,17,20 193:22 201:6 129:21 168:11 331:7 280:14 298:16 128:5 135:16 tuning 182:18 239:5 248:18 transactional 301:10 302:1,4 247:12 249:11 turn 7:18 28:18 251:10 262:17 152:1 320:4 303:7 327:16 265:13 266:13 34:21 36:22 302:10 331:16					
transaction 272:3 274:8 127:12,17,20 193:22 201:6 129:21 168:11 331:7 280:14 298:16 128:5 135:16 tuning 182:18 239:5 248:18 transactional 301:10 302:1,4 247:12 249:11 turn 7:18 28:18 251:10 262:17 152:1 320:4 303:7 327:16 265:13 266:13 34:21 36:22 302:10 331:16					
331:7 280:14 298:16 128:5 135:16 tuning 182:18 239:5 248:18 transactional 301:10 302:1,4 247:12 249:11 turn 7:18 28:18 251:10 262:17 152:1 320:4 303:7 327:16 265:13 266:13 34:21 36:22 302:10 331:16			, ,	, ,	~ <u>~</u>
transactional 301:10 302:1,4 247:12 249:11 turn 7:18 28:18 251:10 262:17 152:1 320:4 303:7 327:16 265:13 266:13 34:21 36:22 302:10 331:16			, ,		
152:1 320:4 303:7 327:16 265:13 266:13 34:21 36:22 302:10 331:16				O	
		· ·			
Haibacuous 357.10,17 300.7 37.7 01.22 370.19					
	น ansacuons	337.10,19	300.7	37.7 01.22	370.19

				. 1
types 45:21	unanimity 97:12	328:10 353:8	universities	usage 38:21
46:12 56:11	unanimous	355:2 360:4	144:4	322:12
85:7 103:11	133:15	367:21 368:2,4	university 2:12	use 18:20 27:15
169:7 219:12	unavoidable	379:7 380:13	2:15 3:3,18	38:14,14 39:10
230:5,5 235:17	360:7	understandable	110:14 117:5	40:13,15 46:5
235:21 236:4	unbelievably	375:11	287:3	52:3 55:13
236:11 268:10	291:1	understanding	unjust 310:22	61:12 63:4,10
typical 70:15	unbiased 378:15	107:9 129:21	unknown 90:5	68:5 69:1
323:1,10	unclear 291:7	239:7 242:2	unlocks 37:7	77:14 78:19
325:15	uncomfortable	244:22 249:1,2	unpleasant 38:2	82:17 86:11
typically 265:3	269:4 363:13	251:10 264:4	unpopular	87:20 91:17
298:10 315:16	363:16 378:22	279:1 291:4	140:18 268:1	94:8 106:5,9
374:15	unconcerned	340:4 344:8	unreasonable	106:11 113:1
typo 137:3	45:6	377:9 380:8	42:3 64:13	115:16 118:17
	undercut 377:2	understands	72:10,13,17	121:10,13,22
U	underlie 55:22	328:8	369:1,4	122:12 123:2
u 3:13 36:4	285:15	understood	unremarkable	123:19 124:3
127:19,20	underlies 28:16	196:10	36:14	125:10 126:1,3
128:4 129:10	underlying	underwent	unrestrained	126:9 131:9
129:14 131:15	28:21 56:6	30:20	298:3	145:19 150:22
131:17 132:1	57:11 84:22	undesirable	unscientific	152:9 155:12
132:12 133:2	85:3 101:14	40:16	17:19	157:11 162:6
164:5 230:2	243:20 333:18	undue 362:8	unsolicited	169:21 170:12
278:22 279:2,3	undermined	unduly 8:8	358:15	170:13 171:3
279:21 289:9	127:12	14:10	unsurprising	174:11,13
292:4 301:8	underneath	unencrypted	315:20	184:10 188:2
302:10 343:17	270:1	162:4	untenable 32:11	198:21 222:9
369:7	underpinning	unfolded 122:19	unusual 186:3	222:15 226:20
uavs 78:20,21	274:14	unfortunately	unwilling	231:18 236:4
ubiquitous	underpinnings	44:13 62:16	108:11	236:11 237:1
50:18 147:19	285:14	unfortunatly	update 168:12	248:16 257:22
148:16 186:5	understand	205:17	updates 374:12	267:10 274:17
379:10	29:1 36:19	unique 34:10,16	upfront 358:9	276:7 288:10
ultimately 23:13	37:20 38:1	39:9 234:10	uphill 243:6	288:13 290:20
74:8 226:19	89:11,12,13	uniquely 205:1	uploaded	291:10 292:2,6
240:10 316:13	157:12 158:2	unit 240:3	139:21 140:2	292:17 294:21
316:20 323:8,9	165:21,22	281:15	upset 265:22	298:11 300:10
372:15,18	178:6 183:14	united 51:8	upsets 302:21	305:22 306:3
umbrella 97:2	195:22 235:3	127:22 193:8	urge 124:8	308:15 309:22
unabated	242:14 245:5	343:22	128:17,19	310:12 312:8
364:11	245:14 247:14	units 228:5	130:14 377:3	312:15 315:19
unable 370:20	249:11,17,18	universal 51:6	urges 303:1	320:10 321:9
unacceptable	252:11 279:12	98:7	usa 212:6	321:12 323:12
271:1	279:13 294:9	universe 298:4	usable 138:16	323:13 324:14
	I	I	I	I

324:19 329:7	29:14 50:22	55:22 56:1,6	98:1,1 132:2	269:9
338:3,5,6	51:4 68:15	57:12 67:6,11	151:22 171:14	voluntarily
341:9,18 342:1	72:9	84:22 97:3	192:6 210:12	219:5
342:6,14	utility 367:11	118:5 209:3	212:12 214:1	voluntary 273:6
344:16 350:10	utilizes 204:20	234:19 331:17	241:1 296:18	voluntary 273.0
351:10 363:17	utilizes 204:20 uttered 66:14	357:18	299:11 322:15	122:8,9
	uttered 00:14	varied 9:4 180:3	324:20 339:21	vote 4:19 384:12
364:8 365:11	$\overline{\mathbf{v}}$			
365:15 372:12	v 133:14	variety 50:16	340:10,11	voyeurism
372:21 380:22	vaguely 269:16	230:16 237:10	342:9 351:6	31:21
381:2,20	vagueness	246:2 247:15	376:8	vulnerabilities
382:16,16	157:17	304:8 308:5	viewing 338:17	49:11
383:12	valid 97:20	various 46:12	370:1	vulnerability
useful 62:1	260:19	76:12 104:14	views 19:11 54:8	16:6
114:6 115:10		105:3 163:22	96:17 231:15	vulnerable
118:8 119:4	validated 227:8	241:7 255:19	284:19 333:17	121:12 123:19
134:5 195:14	valley 135:1	325:18,20	vigorously	
195:16 259:14	136:13	346:1,21 373:1	154:4 339:5	
259:15 273:21	valuable 27:5	vary 95:8	village 76:21	w 1:13
274:1 276:3	93:8 195:6	varying 8:16	violated 29:10	wade 257:20
288:4 289:22	197:12 198:5	196:11	violation 15:10	wait 115:18
293:20 296:15	207:14 291:1	vast 21:14 53:14	281:4,19	201:9 353:18
296:17 332:1	361:6	64:20	violations 15:11	372:11
332:10 333:5	value 9:2 11:4,6	vehicle 40:7	16:22 281:10	waived 122:14
344:22 366:16	15:19 17:20	velocity 50:16	281:11 327:7	wald 2:5 4:15
367:4 368:15	26:11,16,20,22	venture 313:5	violaton 222:20	7:19,20 10:10
381:16	28:15,21 29:14	venue 313:13	violence 160:4	18:1 25:4,9
user 39:15,19	50:19 51:5	verify 247:13	161:22 270:6	33:8,11 41:15
79:17 201:21	54:9 55:12	verizon 293:3	virtually 30:6	43:8 45:10
users 39:11,16	64:4 69:13	version 75:17	virtue 14:2,15	50:10 51:2,15
39:20 112:16	77:6,7 85:3	88:3 100:3,4	visavis 11:1 45:8	51:22 53:2,4
156:8 157:7	98:21 116:19	326:17	188:10,12	53:20 57:14
237:1	117:19,22,22	versus 114:11	vital 214:20	60:21 62:22
uses 15:6 35:4	125:5 132:21	336:5 369:7	379:5	63:19 67:17
71:9 121:15	133:16 144:8	vetting 263:16	vivid 36:15	69:4 71:2
214:19 222:17	214:1,4 222:1	263:22	vocal 127:21	74:10 75:21
234:15 236:7	233:1 244:11	viable 327:5	voice 10:11	80:14,18 87:7
236:12,13	265:19 266:6	victims 113:22	243:3 244:17	88:5 96:6
262:21,22	266:10,17,20	183:15	284:4	102:2 103:9,21
294:13 322:9	270:10,11	video 139:20,21	void 51:14	105:9 106:14
362:7	284:11 291:13	140:2 167:2	volume 37:11	106:18 107:3
usp 164:10	312:5 331:2	329:16	50:15 129:21	108:16,20
usually 39:19	358:8 381:20	view 13:21	228:19 280:9	175:20 177:18
147:16 229:22	values 8:5,15,17	32:10,11 33:15	289:14	178:7 179:12
utilitarian 26:22	26:21 27:8,22	63:2 81:12	volumonous	179:19 180:16
	,	00.2 01.12		

181:3,6 199:9	279:6,8 283:19	197:1 363:12	307:20 309:7	111:20
200:19 201:1,4	287:16,21	wasnt 65:11	327:4 328:22	weakens 166:1
201:9 202:8,20	297:3,11 306:5	181:18 199:22	329:21 330:4,5	weakness 292:3
255:8,12 258:7	309:4 333:13	307:3 358:3,3	332:5 334:21	weaknesses
259:4,7,10,13	335:9 344:1	358:5,6	335:2,8 336:11	342:16
372:8,9 374:6	354:3 356:18	wasted 68:18	336:14 337:12	
377:5,14			338:7 345:3	weapon 111:5
/	360:14 367:22	watch 183:22		weapons 56:22 227:13 365:17
walk 304:21	368:2,4 374:4	198:16 369:17	347:22 348:3	
379:20	374:16 379:15	watched 280:12	351:7,14	wearables
walking 14:2	380:15 382:4,9	watchers 21:13	352:20 353:20	185:17
273:21	wanted 46:13	wave 287:11	354:21 355:10	wearing 164:22
want 6:18,21	47:15 61:16	way 8:10 11:16	355:22 356:9	web 220:10
7:13 20:3,4,4,9	87:14 89:19	17:1 19:3	358:14 359:9	website 7:8 39:9
21:9 22:7 24:6	102:10 142:1	21:16 28:2,19	364:9 373:22	384:8
46:15,18 47:20	180:16 201:10	31:10 32:9	374:4 377:12	wed 146:5,6
47:20 48:7	218:3 220:14	41:2 56:5,9	377:15 378:15	176:13 318:4
51:3 53:8	237:22 262:9	57:11 61:14	380:5 381:16	330:20
63:21 64:8	263:12 273:13	62:12 63:7	382:5,19	weeds 218:5
69:20 76:16	305:13	64:21 67:8	385:10	week 7:8 120:15
83:1,12 86:15	wanting 329:20	78:6 97:22	ways 16:21 21:5	157:8,9 224:13
87:17,19,22	wants 63:3 73:3	98:1,6 113:9	21:17 24:3	373:19
88:4,9 91:1	75:21 106:8	115:4 121:18	26:9 49:19	weekly 184:20
93:14 98:15	264:22 367:18	124:6 125:18	68:1,16 69:16	weigh 52:5,22
99:3 104:16	379:19	144:12 145:4	74:18,21 89:6	53:6 66:18,20
108:1 115:2	war 122:6,7,15	145:10 150:11	99:15 116:11	67:1,5
120:3,13 133:4	122:18 187:6	164:17 172:22	147:16 148:11	weighing 33:5
143:21 146:21	228:2	173:2,11 175:6	149:14 163:9	232:22
147:1 156:9	warning 110:2	175:22 183:6	164:11 165:9	welcome 4:2
158:22 160:4	warrant 29:18	190:2 191:20	184:16 191:15	7:10 134:18
169:9 181:14	53:10,16 54:7	192:4 197:6	199:5 214:3	162:11 181:13
185:10 189:9	65:11,13 70:18	198:16 200:4	215:2 240:17	203:9,10 225:7
192:5 194:2	71:21 73:1	200:16 201:17	240:18 244:20	237:14 375:22
198:3 202:2	81:21 132:11	201:21 202:6	258:4 261:2	376:1
206:22 207:1,7	167:22 168:6	202:13 209:1	265:14,20	welldefined
207:17 215:5	168:10,21	210:21 211:5	277:20 279:7	155:2 158:13
235:20 239:2	169:5	224:19 229:6	280:17 295:7	wellunderstood
243:22 248:11	warranted	230:12 242:5	307:22 319:5	14:6
250:7 257:12	366:2	258:12 260:6	334:6 337:17	wellwritten
263:4 265:5,18	warrants 42:4	260:12 261:3,9	339:3 350:12	351:19
267:11 269:3	70:11	265:12 271:20	371:4 373:13	went 45:13
269:18 270:17	wartime 123:11	276:7 283:20	374:21 379:20	80:19 160:16
270:21 271:8	washington	285:12 287:11	weak 24:11	161:11 175:3
273:15 275:16	1:13 2:15 4:7,8	289:7,22	weaken 111:21	175:12 304:20
275:21 278:6	18:4 130:19	294:22 304:14	weakening	315:2 374:11
			ľ	l

westend 4:6	292:7	women 35:21	328:14 329:14	312:16 331:8
westeria 4.5 weve 15:12 32:9	wherewithal	270:6	349:19 356:17	331:12 335:19
51:13 58:6	195:3	won 278:14	356:18 371:20	339:11 356:22
92:3 95:18	whispers 164:19	wonderful 99:8	373:19	361:17,19
126:16 127:4	wholesale	143:6 167:6	worked 127:12	363:18
120.10 127.4	108:12	wondering 55:1	136:5 156:18	worried 178:3
128:4 129:4,13	wholl 367:20	82:11 259:21	176:17 226:1	258:13 266:18
129:18 131:7	wholly 315:10	271:21	289:9 303:18	304:22
	316:11			
132:11,15		wont 15:20,21	worker 59:22	worry 15:14
150:3 154:17	whos 60:4 175:4	105:21 191:17	workforce 45:4	18:10 22:17,18
168:4 169:5	196:1 220:10	366:15,19	171:19,20	43:15 68:11
172:9 173:6	250:19 256:14	word 27:7 31:12	172:11 214:16	120:16 147:7
203:13,19	285:1 328:18	71:2 248:15	working 12:11	161:13 176:7
229:5 232:13	wide 282:19	words 66:14	144:3 145:20	349:9
235:19 262:11	319:1 364:12	108:18 211:8	205:9 213:16	worrying 20:10
262:14 263:1	widely 204:8	218:11 240:11	239:6 242:12	worse 138:6
269:2 284:3	283:3 337:15	256:20 257:3	242:17 250:3	worst 13:20
285:12 320:18	willful 95:9	262:18 329:13	250:20,22	16:21
331:13 337:16	222:11	330:8 331:2	251:12,13	worth 26:22
337:17 341:18	willing 8:16	350:20 355:11	252:5 275:8	227:6
343:9 349:9	13:18 41:11	work 38:1 62:19	283:12 311:7	worthless
350:22 366:21	85:17 87:13	88:21 91:15	312:18 347:5	117:20
367:1 373:1,17	101:5 106:9,10	95:3 99:15	workings 129:7	wouldnt 64:5
383:7,8,11	282:12	100:15,15	works 54:12	82:19 149:18
whack 63:20	willingness 95:6	132:22 133:2	83:10 89:14	180:11 202:9
64:16	willynilly	140:6 143:9,9	153:16 189:18	307:6 318:15
whacks 64:18	107:14	145:17 148:19	227:14 251:22	355:18 380:15
whatnot 247:10	wilson 7:14	148:20 154:10	251:22 295:11	381:4
whats 23:7	wind 346:17	182:10 189:14	304:5 356:10	wrap 9:13 96:10
24:21,22 25:1	winn 6:15 7:16	189:19 195:20	world 13:10,19	134:8
28:14 51:15	winning 8:6	200:14 205:15	28:12 32:12	wrapped 59:15
75:17 76:6	wins 19:18	207:10 217:6	55:15 110:21	wrapping 207:5
78:4 81:12	wiretap 166:11	224:7 234:12	122:7,18	wrestling
84:9 93:12	wish 135:5	238:17 240:5	131:21 132:2	368:17
98:11 123:18	177:21	241:3 246:7,11	133:9 135:19	writ 66:20 67:5
126:16 160:8	withhold 14:9	246:20 255:14	135:21 136:16	67:6 333:9
176:14 177:15	withstand 101:2	270:1 273:15	136:20 137:5,5	write 88:10
192:18 244:22	112:14	273:17 274:15	137:22 140:9	136:20 145:9
250:21 251:5	witness 117:3	283:15 284:8	157:22 140:9	286:18 318:11
	124:11 385:12	289:12 296:1		378:18
256:19 326:4			166:5 187:6	
330:21,21	witnesses 96:9	308:11 314:3	190:4,5,9,15	writer 102:9
364:21 379:21	96:19 99:21	317:11 319:2	191:9,16 198:1	104:5
380:2	109:16,19	319:13 327:1	237:8 250:22	writers 16:13
whatsoever	woman 37:8	327:12 328:7	289:13 304:8	writing 16:15
	1	I	I	l

83:2 312:12	year 7:12 128:2	166:21 182:14	1	2014 1:7 4:5,10
writings 61:4	129:10,15	194:2 199:7	12:97:5,20	16:8 75:17
written 6:14 7:9	130:19 217:20	212:19 220:2	203:1 352:11	385:14,19
9:20 30:15	232:14 284:3	241:21 258:13	10 109:2,3	215 30:17 60:14
64:3 118:19	384:6	259:2 260:8,13	284:17 379:13	90:1,9,14 92:3
119:6 121:7	years 10:16 94:9	261:20 265:18		103:5 128:16
144:6 145:4	123:14 124:13	265:19 270:2	385:18	175:10 181:1
146:22 195:13	125:2 128:6	272:17 273:8	100 352:8,11	201:5 257:3
223:13 240:21	137:15,16	276:14 277:22	11 217:12	278:20 279:9
276:2 318:10	149:6,19	279:2 283:5	12 1:7	301:4,11,16
318:22 319:3	153:20 157:3	287:10 305:1	1200s 76:22	314:20 330:11
326:13 328:5	163:13 167:10	310:15 312:10	1221 1:13	351:17 354:3,4
328:18 384:4	177:2,2 181:22	315:17 318:14	12333 82:6	360:14 365:9
wrong 71:16	227:19 229:9	320:12 323:17	164:7,10	369:5
93:16 107:16	230:15 288:6	332:3 338:15	229:18 271:14	21st 4:10
166:20 193:9	306:12 311:20	347:9,11	278:22 301:5	22nd 1:13
294:20 322:4	322:9,10	348:16 349:16	364:17	25 229:9
328:20 330:7	360:18 361:2	349:20 350:10	12th 4:5	27 141:8
331:21 353:9	379:13	351:2 353:18	14 338:19	2703 169:3
369:22	yellow 9:11	357:18,20	15 7:5 109:2	29th 137:4
wrongdoing	110:1 207:5	367:4 371:12	167:10 203:1	29th 137.4
281:2	286:4,5 293:15	378:9	230:15 284:18	3
wrongful	293:17 370:12	youtube 139:21	379:13 384:14	3 3:9 88:18
222:21		139:22 140:2	1942 122:13	110:10 141:8
•	yep 277:6,8		1970 s 25:21	30 1:14 4:5
wrongly 52:4 wrote 45:12	yesterdays 123:22	youve 16:4 30:15 61:3	73:22 210:9	109:3 163:19
46:8 119:12	yield 112:13	105:15 108:20	1973 26:2 40:5	315 361:1
288:7	189:8	120:14 149:7	119:12	
200:7		248:22 249:22	1980s 329:4	4
X	york 241:10	273:1 289:4	1989 14:12	4 3:16 384:14
$\frac{1}{x}$ 183:22,22	youd 255:4 307:2 318:9		2	40 224:8
184:1,3,4		297:10 313:13		45 284:18
363:14	youll 41:4 69:1	322:1 353:9	2 3:1 284:18	47 16:9
xyz 202:1	69:2 177:2 271:7 286:5	354:10 383:16	303:19	484 104:8
Ay 2 202.1		$\overline{\mathbf{z}}$	20 9:16,17 41:16	4g 136:22
<u> </u>	287:11 293:17 335:16	z 363:14	74:11 141:19	
y 363:14		zero 151:3	157:3 286:9	5
yall 238:1,2	young 17:7,8,9	zeroday 111:19	2005 181:15	5 110:10
yeah 136:9,11	17:19 younger 338:16	zones 40:9	2007 123:11	50 70:12 157:9
147:14 157:11	younger 338:16 youre 68:14	zoom 170:19,19	137:4 158:12 2008 100:5	157:10 306:12
162:20 168:3	69:18 77:19	170:19		363:13
168:22 187:21	83:13 99:12		217:18 271:16	51 196:22
188:1 200:20	100:18 150:19	0	2011 123:13	520 16:12
281:9 339:12	152:13,22	0 303:19	2013 141:7,10	
	134.13,44			

		/(
60 39:18 329:14 62 227:19 622 17:11		
622 17:11 623 17:12 7 700 220:8 702 31:3 163:11		
163:21 164:5 174:3 213:19 256:9 278:22 279:10,19,20		
300:11 358:1 74 163:1 75 39:18 163:1 8 8 1:14 4:5		
9 9 217:12 99 352:8		