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Department of Homeland Security Meeting
Implementing Privacy Protections in
Government Data Mining

International Ballroom East

Hilton, Washington

July 25, 2008

Attendance: Peter Sand, Michael A. Aisenberg,
Brian Tretick, Daniel Weitzner, K. Waterman, Toby Milgrom
Levin, Fred H. Cate, Thomas Oscherwitz, Barry Steinhardt,
Peter Swire.

Transcribed by:

Dana A. Cohen

For

Alderson Reporting

1 [Convened at 8:38 a.m.]

2 Mr. Peter Sand: Thank you very much. Brian, do
3 you want to start us off?

4 STATEMENT OF BRIAN TRETICK, EXECUTIVE DIRECTOR
5 FOR PRIVACY ADVISORY SERVICES, ERNST & YOUNG

6 Mr. Brian Tretick: Good morning, again. I'm
7 going to read from some prepared remarks, so I make sure I
8 cover the points that we had discussed in preparing for
9 this panel. I'm going to take off my glasses so I can see
10 my paper. I can't see you --

11 [Laughter.]

12 Mr. Brian Tretick: -- if anybody will interrupt,
13 you need to speak up.

14 So, again, I want to continue the conversation
15 with a discussion of basically this thing called the
16 "audit" and I'm using air quotes around it, and a lower-
17 cased A for audit, because it means a lot of different
18 things to a lot of different people, and most of those
19 people don't use it -- the term audit -- with any kind of
20 precision.

21 So, let's focus on the definition from the
22 American Institute of Certified Public Accountants, I don't

1 know that we have any other members of that august body in
2 this room, but me, but the AICPA publishes -- one of the
3 things it does, it publishes professional standards for
4 auditing and accounting. AU 100 is the foundation for the
5 financial audit, the legacy from which audits over other
6 subject matter, non-financial subject matter, are based.

7 So, I'm going to paraphrase from its opening to
8 just describe, again, the philosophy behind an audit, okay?
9 The objective of an ordinary audit, and this is of
10 financial statements, by the independent auditor, is the
11 expression of an opinion -- so the objective is the
12 expression of an opinion, in this case of the fairness with
13 which the financial statements presents in material
14 respects, the position of the company financial, its
15 operations, its cash flow.

16 So, again, let me summarize that. The opinion --
17 the objective of the audit is the
18 expression of an opinion about the reasonableness in
19 material ways, okay? So, that's what we want to get at.
20 So, it's not absolutism, and it's not on the immaterial
21 side. So, again, a company may have had some errors in its
22 tallying of its financials, but as long as those aren't

1 significant and material, we look for issues of the
2 fairness of the representation of the financial position of
3 the company.

4 So, that's a philosophical aspect of it. So, the
5 standards -- so the auditor looks at the subject matter and
6 applies something called generally accepted auditing
7 standards. You know, this term "generally accepted," we're
8 going to hear over and over again.

9 And really, what that means is, a bunch of
10 auditors got together, and floated a standard, we know the
11 standard process in a number of areas, and then at some
12 point it gets pushed out for scrutiny -- these accounting
13 practices, and auditing standards have been around for
14 decades, and evolved over those decades. But -- so, that's
15 where we get the "generally accepted" part.

16 The auditor has the responsibility to plan and
17 perform the audit, to obtain that reasonable level of
18 assurance, about whether -- in this case the financial
19 statements are free and material, and the statement,
20 whether they're caused by error or fraud, it doesn't
21 matter.

22 So, the important point here is that the audit is

1 designed and executed, again, to obtain reasonable
2 assurance -- not absolute assurance -- typically in the
3 financial community, and this is what's acceptable, focused
4 on material misstatements. This reasonable assurance,
5 then, is the basis of an auditor's opinion. Okay, that's a
6 simplification -- there are hundreds of pages in this AU
7 100, just in the audit guide itself -- but there are AU
8 series documents, or AT series documents, more than you'll
9 ever want to imagine.

10 Traditional audit standards have been augmented
11 more recently in the last couple of decades, with standards
12 for more general attestations. But they carry forward the
13 principles of independence, objectivity, completeness from
14 the auditing standards.

15 Under ATIS Standards -- and these are general
16 non-financial audit standards. An auditor may express an
17 opinion about things in other subject matters, such as
18 measurable characteristics. One of the things we might do
19 is the physical size of a factory, or real estate or
20 something, or the number of bricks in a brickyard, you
21 know, there's some physical aspects.

22 Historical events, such as the price I offered to

1 customers over certain dates, might be something somebody
2 wants to know. Or even compliance with laws and
3 regulations, and more importantly for this subject,
4 internal controls over some sort of process.

5 So, the last two -- compliance with laws,
6 regulations and other obligations, and internal controls
7 are the most important for us, here.

8 So, let's come back to the topic, audit over an
9 organization engaged in data mining, might be designed to
10 provide insurance, okay, independently and objectively, so
11 that the so-called internal controls over that process --
12 data mining, data sharing, whatever -- whatever you want to
13 call it. The controls are in place, designed
14 appropriately, so not missing, and they're pretty good --
15 they're good at the -- they're the right level of rigor.
16 Okay, in place, designed appropriately, and operating
17 effectively over time, okay? So, that's the controls part:
18 in place, designed appropriately, and operating effectively
19 over time.

20 The other component to this that we'll want to
21 consider here is that it complies with the laws,
22 regulations, or other obligations that are imposed upon it.

1 The rules, for example, data usage scenario that you've
2 discussed. So, these -- this is the basis that we want to
3 put forth -- an internal controls audit, and a compliance-
4 type audit could apply here.

5 So, who audits is a second question that we're
6 asked to discuss. The answer depends on the purpose of the
7 audit, the nature of the assurance that is required. In
8 some cases, the management -- the organization itself --
9 may use its own organization, perhaps its internal audit
10 function, or its own business team, its own group to
11 conduct the audit. And others, you may like to even have
12 or be imposed an external auditor, or Inspector General,
13 for example, over that process to conduct the audit.

14 The assurance that you can derive from the audit
15 often depends on the selection of the auditor. Internal
16 self-checking may be appropriate when, for example, the
17 audience is internal, okay? So -- and you can rely upon
18 that auditor.

19 An external auditor, an independent external
20 auditor, would be most appropriate for developing assurance
21 when it needs to be conveyed to the outside parties.
22 Again, the reason you bring in an auditor is to develop a

1 position of assurance, reasonable assurance over something.
2 But you have to be able to believe the messenger. Again,
3 my mother -- my mother loves me, she's not objective. So,
4 if you're looking for a character reference for me, don't
5 ask her. But, you'll want to -- if you really need to rely
6 upon it, you'll want to get character references from
7 external parties. So, it's the same kind of argument in
8 this case.

9 So, independent auditor is, again, a common
10 condition of SEC-mandated financial audits, and even the
11 FISMA audits the government agencies go through.

12 So, the AICPA, the American Institute of
13 Certified Public Accountants, got together in 2001, and by 2003
14 came out of a conference room and published what has become
15 known as generally accepted privacy principles, there's a
16 reference on my last slide, to where that is located. This
17 body of work translated the ATIS Standards for the subject
18 of controls over the use of personal information, so
19 privacy audits can be done, and basically addressed the
20 common principles for privacy and data protection that were
21 prevailing, and would have included U.S. Federal government
22 rule -- common U.S. Federal government rules at the time.

1 It's like the garlic in the Ragu sauce, "It's in there."

2 It's baked into that process.

3 So, basically, the result of examination
4 procedures, planned and conducted by the independent
5 auditor, in this case, the audit, would be an opinion over
6 what the organization claimed it did, its policies, its
7 privacy policies and practices, and over the organization's
8 ability to meet those criteria. So, the generally accepted
9 privacy principles gave us a foundation, a common
10 foundation, in which we as an industry could conduct
11 objective, relevant, complete, and measurable privacy
12 audits.

13 More recently, as an example, the Transportation
14 Security Administration established its own requirements
15 for security and privacy, within the commercial Registered
16 Traveler program, and requires participating organizations
17 to submit reports of their independent auditors, according
18 to those requirements on an annual basis -- the reports
19 convey the opinion of the auditor, that it, the auditor, has
20 developed the position of reasonable assurance that
21 personal information was used, stored and retained, in
22 conformity with the program's requirements. So, the TSA

1 took a bunch of the NIST requirements, and then added
2 things that were specific to the Registered Traveler
3 Program, to create the security and privacy standards that
4 it has to meet.

5 One of the Registered Traveler Programs, in
6 particular, Verified Identity Pass, which operates the
7 CLEAR program, you see it at Reagan National and Dulles as
8 an example, they also had their system audited against the
9 AICPA standards, so they can issue a public report to their
10 members. So, they're not only providing assurance to the
11 TSA as required by assurance to their members.

12 Finally, what gets done in audit, remember, it's
13 up to the auditor, normally, to determine the procedures to
14 be executed, and the auditor is the one who issues the
15 opinion, so we're responsible for coming up with the right
16 procedures to develop that position of reasonable
17 assurance, because it's our signature at the bottom of the
18 report, and our reputation, and in many cases, our
19 liability.

20 The generally accepted privacy principles offers
21 an auditor illustrative controls, things to look at.
22 Additional considerations, plus, again I want to refer to

1 you to the thousands of pages of audit and ATIS Standards
2 that come along in the backpack of any auditor who graces
3 your door, so it's not a crap shoot, in any way.

4 I think I want to flip on -- so basically to cut
5 this short, observation, inquiry, inspection and test --
6 these are all common test procedures, these are the tools
7 of the auditor, as well. I want to test the -- whether the
8 -- the rule that the organization has set up, I want to
9 test whether there's a controlling place, it's designed
10 appropriately, and it's operating effectively over time,
11 and this is what gives me the basis for my reasonable
12 assurance.

13 The aspects of generally accepted privacy
14 principles include things like the management of privacy
15 and security policies, procedures, programs and controls,
16 so the administrative part of it. The things affecting the
17 individual; the notice, the choice and consent, the redress
18 programs and other processes that are focused on the
19 individual, the subject of the data, if that's applicable
20 in this case.

21 Controls over the life cycle of the personal
22 information, that's the collection, use, retention and

1 disclosure I mentioned earlier; disclosure to other parties
2 involved.

3 Security, integrity, quality and other similar
4 requirements, and then finally, the monitoring of controls,
5 because an auditor can't have confidence that controls are
6 in place and operating effectively just by looking at
7 themselves, they need to also see that the organization has
8 an ongoing program of monitoring the effectiveness of their
9 own controls. Without that, it's difficult for me to form
10 a basis of reasonable assurance over your operations.

11 So, in most organizations, use of personal
12 information is complex, and not only that, especially in
13 the case of the subject of the hands, you know, these
14 arrangements for data sharing, data mining, are going to
15 cross organizational borders internally, and externally.
16 So, it complicates this thing called the audit.

17 There are legacy approaches to auditing such
18 complex environments. First thing we do is we take a
19 system like this and we decompose it, okay? Where the
20 requirements, then, map out all of my requirements on my
21 claims that I want to make, or the rules of the law,
22 regulations or other agreements, and say, "Where are they

1 applicable?" Okay? Because, they're not uniformly
2 applicable in all entities and all processes, okay? So,
3 the first part is the decomposition to the processes that
4 handle personal information, and the supporting, as we'll
5 hear later, technologies and manual procedures, okay? So,
6 some of that requirements that we have at the highest level
7 aren't applicable in some of these areas, so we want to
8 rule those out, and that's okay to do.

9 The decomposition shown here includes multiple
10 parties under different controls and governance, a common
11 audit strategy may involve either the auditor going to
12 each one of these boxes on this chart, and conducting
13 examination procedures. Or as another option, I can look
14 at the third party on the right side, and say, "I'm not
15 going to come and audit you, but you need to provide me a
16 reasonable basis of trust and confidence in yourself."

17 And this is, again, for example, looking back at
18 the TSA Registered Traveler Program, the TSA is not going
19 out and auditing everybody. But they're asking for, again,
20 under a certain program, an audit report from these
21 organizations, so that you can play within its cloud, okay?

22 This is the same kind of basis of assurance that

1 we have in the financial community where, if I'm a
2 corporation, I have outside service providers -- payroll
3 providers, other people processing my financial information
4 -- I don't want to send my auditors everywhere to audit
5 every one, and these service providers don't want to be
6 audited by all of their customers.

7 So there's a construct, in this case, a SAS 70
8 Reports -- Statement of Auditing Standards number 70,
9 again, they go up into the hundreds, so beware -- but a SAS
10 70 Report allows an organization to be audited once, and
11 that the audit report is an auditor-to-auditor report, it's
12 really big and ugly, but what it tells any other people who
13 come knocking on the door, it says, "Listen, we just got
14 -- we just got audited, here are the results," and an
15 auditor, again, if that's done under generally accepted
16 auditing standards, I can rely on somebody else's work in
17 that case, okay? So, we look at the reliance on other
18 parties' work.

19 Okay, so I think what I'd like to do is kind of
20 cut to the chase. When looking -- when you are looking at
21 the philosophy, the objectives, the methodology and
22 practices for privacy audits in data mining, the Department

1 and the other parties here should take into consideration
2 the legacy of this financial audit; more importantly, the
3 interpretation by the AICPA in the generally accepted
4 privacy principles.

5 Where audits are needed, the Department can look
6 at the recent example of the Transportation Security
7 Administration as a, again, a worked example in this case,
8 although, again, it's not data mining.

9 And with that, I'd like to thank you and turn it
10 back over to Peter.

11 [Applause.]

12 Mr. Peter Sand: Thank you very much, Brian.

13 STATEMENT OF DANIEL WEITZNER, CO-DIRECTOR,
14 COMPUTER SCIENCE AND ARTIFICIAL INTELLIGENCE LABORATORY
15 (CSAIL) DECENTRALIZED INFORMATION GROUP, MASSACHUSETTS
16 INSTITUTE OF TECHNOLOGY

17 Mr. Daniel Weitzner: Hi, everyone.

18 Thanks very much to Pete and Toby and Martha for
19 giving us the opportunity to come and talk with you. As I
20 said, my name is Danny Weitzner, and I run a research group
21 at MIT, where we look, in general, at advanced web
22 technologies and the public policy implications of many of

1 those technologies.

2 So, we have a project that we've been working on
3 for about the last two and a half, three years, called
4 Information Accountability, and our interest, really, is to
5 understand how to design large-scale information systems
6 -- whether they're data mining systems, or the web, or
7 large enterprise systems -- how to design these very large-
8 scale systems in a way that they have a property that we
9 call information accountability.

10 What we mean by information accountability is
11 that it's possible to keep track of how information is used
12 in an information system, and then where there are misuses,
13 to detect that misuse, and hold accountable either the
14 individual or the institution, as appropriate, for that
15 misuse.

16 I want to talk, quickly, about some of our larger
17 motivations for doing this work, and I'm -- we're going to
18 come back to the specific application to data mining
19 context, but I want to talk about the bigger picture.

20 As we've looked at, and been involved with, the
21 design of much of the next-generation web technology, what
22 we've seen -- both in the technology and in the way people

1 use it -- is that we're really living in an increasingly
2 transparent environment. It's not a value judgment, it's
3 just a statement of how we all interact with personal
4 information, and information, in general.

5 The picture you have here -- I'm going to tell
6 you just a little story to illustrate the kinds of problems
7 we're interested in solving -- the picture you have there
8 is a diagram of a part of a social network -- it's actually
9 the social network at MIT, and some students of mine
10 collected some information from that social network, and
11 set out to test the hypothesis whether they could discover
12 in this social network who was gay, and who wasn't, based
13 on a very small amount of information. They happened to
14 get this data from Facebook, and we can talk separately
15 about how that worked.

16 [Laughter.]

17 Mr. Daniel Weitzner: But, the important thing is
18 that, simply by looking at information that people
19 routinely disclose in the course of their interaction in
20 that part of the web, in that kind of system, it ended up
21 that people were revealing, really, quite a large amount of
22 information, in some cases information they didn't quite

1 intend.

2 And this is kind of a -- I think just the -- sort
3 of one indication that, in some respects, we're really all
4 data miners, we really all -- have the ability in one
5 way or the other to learn quite a lot about people out on
6 the web. You'll know, certainly, about the -- the more
7 controversial data mining programs, I think you probably
8 know that the person who designed this logo was asked to
9 retire, at some point, but I think that what we do know is
10 that there really are many socially important uses --
11 whether it's -- of data mining -- whether it's for national
12 security purposes, or for learning more about our health
13 status, for getting more personalized treatment of our
14 health conditions, for sharing information. We're all in a
15 situation where we're going to be making more and more use
16 and sharing information more and more about ourselves.

17 And so the question that we've come to is to ask,
18 once lots of information is disclosed and available about
19 people, what can we do about privacy protection? Do we
20 have to just say, "Forget it, get over it, it's out there,
21 you have no more privacy," or do we have a way to
22 reintroduce some set of controls on how that information

1 can be used?

2 So, again, our research is really focused on the
3 question of -- once information is flowing rather freely in
4 any given system, can we still maintain control over how
5 that's used, and have some rules about what are acceptable
6 and unacceptable limits on personal information.

7 In the context of the technical projects that
8 we're developing, looking at data mining models, we have
9 really three very particular goals, and my colleague, K.
10 Waterman, will show you some of how we've implemented them.

11 First and foremost, what we're trying to do is --
12 while we want to enable data mining to go on, we want to
13 prevent mission creep where possible, we want to make sure
14 that once information is collected and analyzed, that it's
15 used for the purposes that were intended, that it follows
16 whatever the rules are about how that information can be
17 used.

18 At the same time, we want to enable more seamless
19 sharing of information. Our sense -- and I think you
20 probably know this far better than we do -- is that in many
21 cases information sharing and analysis that actually would
22 be permissible under the rules, doesn't happen because it's

1 hard to tell whether that analysis is actually allowed, or
2 whether that usage is actually allowed or not. So, we want
3 to put a wall around mission creep, but enable the kind of
4 sharing that's actually possible.

5 The way that we approach this is to implement a
6 system-wide and very fine-grained audit, both of what
7 information is accessed by whom, and most importantly, how
8 that information is used, and whether those uses conform to
9 rules.

10 Just very quickly, this is research that's going
11 on with a team that we've put together that includes
12 colleagues at MIT, Computer Sciences Program at Yale and
13 RPI, and I'll -- we're going to try to stay away from the
14 technical details here, but what I do want to do is, I'm
15 going to turn the floor over to K. Waterman, whose going to
16 talk about one particular application that we've developed
17 to illustrate how this technology works. And then I'll be
18 back.

19 Ms. K. Waterman: Great. So, we like to work
20 with hypotheticals that are based in fact, so in this case,
21 we're looking at a public health investigation. The CDC is
22 investigating a case of extra drug-resistant tuberculosis

1 which is a real problem and, actually, came up after we
2 started working on this scenario, but basically it's a form
3 of tuberculosis that's incredibly hard to treat, and so it
4 is important for them to quickly figure out who the patient
5 has been in contact with, and begin to work with those
6 people to figure out if they also have the disease, or to
7 do what they can to stop the spread.

8 In our hypothetical, so we could get to data
9 mining, our patient is in a coma, so they can't ask that
10 person any question, and they have to use data mining to
11 try to figure out who this person has been in contact with.

12 They use traditional data mining techniques, and
13 one of them is to go to the phone company, find out who
14 this person has been calling, and then look at those phone
15 records to try to figure out how close the association is.

16 Later, a person who was one of those people whose
17 phone records was checked, goes to get a phone company to
18 send out a service guy, a repair guy, and the phone company
19 refuses.

20 So, this -- that's our hypothetical, we're trying
21 to figure out whether that refusal was permissible.

22 So, the way that we do that is, right now we're

1 working with the user interface where, on the left-hand
2 side -- I'm not sure I can -- so, on the left-hand side we
3 have a little sidebar, and basically what it does is it
4 says, first, what transaction log do you want to look at?
5 What -- where is the file where the system recorded what
6 happened? Somebody entered one, pressed the button, and on
7 the right-hand side what you see is our representation of
8 what those events were that would have been recorded in a
9 systems log -- somebody pressed this, somebody requested a
10 copy of that, that sort of thing.

11 The next thing that happens -- let's see, is the
12 system says, "And what policy do you want to know about?"
13 Basically, what rule are you asking if this series of
14 events complied with? So, a little further down in that
15 same thing is another little box like that where they put
16 in where the file is that has the representation of the
17 policy, they press the button, and the right-hand side
18 fills with our form of representation. Understandably,
19 this is a little more to the programmer's side than to the
20 end-user's side, but sort of pretty far down there, it's
21 breaking the rule into a pattern, and at the end there's a
22 place where it's showing that refusing a request for public

1 service based on health information is not permissible.

2 So, then it actually puts those two things
3 together -- it reasons, using the rule, over the data it
4 has, and it comes to a conclusion. And the first one,
5 again, is sort of a little programmer-y for most of our
6 folks, so we said, "Okay, let's do something that's human
7 readable."

8 And one of my favorite things, because I was a
9 five-year-old who said the question, "why?" probably more
10 than anything else, is we have the little "why" button.
11 So, the system says to you that this isn't compliant,
12 there's a little sentence down here where it says to you
13 that the -- Betty, who's our operator, rejecting Bob's
14 request -- is not compliant with the Massachusetts
15 Disability Discrimination Policy, and then you get to press
16 the little "why" button, and ask it, "Well, how come?" And
17 it begins to give you more information and you can just
18 keep pressing, why, why, why, and it will give you more
19 information.

20 But, we still said, that might not make people
21 completely happy, so we're now playing with a sort of
22 lawyer interface, to show you what the system is really

1 doing underneath. Because it has the transactions, and the
2 rules, it can actually put together what those of you who
3 are lawyers may remember from first year of law school -- a
4 traditional legal reasoning. It will basically say to you,
5 "What's the issue?"

6 And this issue -- it's putting this sentence
7 together based on the data that it has. It's always asking
8 whether some transaction log is compliant with some rule,
9 so here it's the transaction log, is it complying with the
10 Massachusetts Discrimination Policy, because that's what
11 somebody asked.

12 And then it says, "Well, what rule?" Well, the
13 policy that you told it to look at is one that says that to
14 be compliant, denial -- the Denial-of-Service rule part of
15 the Massachusetts Disability Rule says that the reason for
16 the event has to be something other than a disability, it
17 can't be looking for a disability.

18 It goes on, and it says, "Well, what do I know
19 about what happened here? What are the facts of my case?"
20 And it says Bob Same, who's our phone company customer,
21 made the request, and it was refused because of a
22 particular record, that the system shows it's one record

1 where the refusal took place, and that this man is a
2 resident of Massachusetts, and we know that he's, in fact,
3 covered by this law, so then it does the analysis.

4 And it says, "That request was denied based on
5 health information," the category of the information was
6 health information, and that's what's contained in that
7 phone record -- the phone company record they were using,
8 and so it ultimately concludes that the transaction of
9 Betty rejecting Bob's request, isn't compliant, is non-
10 compliant with the Disability Discrimination policy.

11 So, what it's doing, just to repeat, because
12 that's a lot, fast, is it's taking a system log of
13 transactions that have occurred over time, and it's running
14 a particular rule or policy that you asked it about, over
15 that transaction log to see if it was compliant or non-
16 compliant with the rule. And, of course, as you can
17 imagine, you could do that with lots of other data mining
18 questions.

19 And with that, I'm going to turn it back to Danny
20 for, perhaps, a little more technical explanation.

21 Mr. Daniel Weitzner: Thanks. I'm going to talk
22 really briefly about how we do this. I should say, first

1 of all, as to some of these user interfaces, we think we've
2 done a lot of interesting work here, technically, we don't
3 think the user interfaces reflect the state of the art of
4 anything in particular, so this is a -- just as a matter of
5 mechanics, this is a little extension that we've written
6 for the Firefox browser, at the end of the slides you'll
7 see there is a place you can actually download this stuff
8 and try it yourself, but an operational system would have
9 to do a somewhat better job.

10 What's -- what we've done that I think will have
11 some value, as K. said, we have a general mechanism by
12 which we can look at records of events, records of events
13 of information usage and access, and then a way to record
14 what the policies are that govern that usage and that
15 access, and come up with results like this. So, this -- we
16 spent a lot of time putting together -- mocking up the data
17 that reflects this example, that's reflected in this
18 example, but the underlying mechanism is, we believe,
19 relatively generic, so it could be used in lots of
20 different contexts, and in fact, could be used for rules
21 that even have nothing to do with privacy. You might use
22 it for rules that assess the reliability of information or

1 assess who's allowed to use different information based on
2 its classification status, or anything else.

3 Very quickly, the system that we've built -- and
4 I'll stress that this is a prototype, proof-of-concept
5 system, has three significant components. First of all,
6 there is a mechanism for doing relatively traditional
7 access control, through rules-based access control
8 techniques. That's not really new, but we've integrated
9 that into the existing rules and policy description
10 language that we have.

11 All of the access events and the usage events in
12 the system are then logged in the system. What we're
13 working on right now are techniques by which we could
14 analyze logs that are generated by the systems that you
15 might use -- the enterprise platforms that are commonly
16 used in government and private sector, and transform those
17 logs, of which as you know, there are many -- transform
18 those logs into data that we can actually feed into a
19 system like this, and get some accountability answers.

20 And finally, we've designed all this, really, as
21 a set of components that can be used in the context of the
22 worldwide web, so these are components that can be added

1 onto browsers and web servers. What that means, we
2 believe, is that this is an architecture that can be
3 relatively easily implemented in the context of any
4 reasonably open architecture -- we're looking now at ways
5 to integrate these tools into enterprise architectures.

6 A little more about the approach that we've
7 taken, and this is where I will try not to fall off a
8 technical cliff -- I'd say that what's important about what
9 we've done here, the technologies that we've used -- as
10 distinct from other kinds of access control systems and
11 production rule systems that are widely used, is that we've
12 implemented a technique which is known as dependency
13 tracking, in the reasoning. That is, the computer program
14 that takes the logs that you've heard about, and the rules
15 that you've heard about and puts them together to try to
16 get answers about compliance, uses a particular style of
17 reasoning.

18 And the important thing that it does is that, it
19 not only gives you an answer, a yes or no answer, "Are you
20 compliant?" or "Are you not compliant?" but it also gives
21 you an explanation for that answer -- it allows you to
22 actually look through the logical process that the reasoner

1 went through to help you understand why you either are
2 compliant, or are not compliant.

3 We didn't show it, here, but one of the useful
4 features of this kind of dependency tracking reasoning, is
5 that in many cases, as you well know, you don't get a
6 simply yes or no answer to a question about whether a
7 particular information usage, or information is compliant.
8 What our reasoner will allow you to do, is to get as much
9 information as the program can glean about whether a
10 particular event is or is not compliant, and will then say,
11 this is dependent upon a certain set of conditions or on a
12 certain set of facts, and if you believe those facts or
13 believe those conditions, then you're compliant. And if
14 you don't, you don't.

15 So, what we're really trying to do is to go for a
16 kind of a hybrid system that allows for integration of as
17 much machine judgment as possible, but in the end,
18 application of human analysis of what these results are.

19 I want to just close by giving my opinion about
20 how this -- how we ought to think about these kinds of
21 information accountability tools in the larger privacy
22 context.

1 We think that these tools will be necessary in
2 order to have a reasonable kind of compliance in complex
3 data mining environments. We think that it is going to be,
4 because of the scale of information analysis that goes on,
5 because of the variety of different rules that may go into
6 any given analytic effort, we think that some amount of
7 machine assistance is needed, in order to supplement human
8 judgment about compliance. But we don't think that these
9 kinds of systems replace either procedural rules, or
10 substantive legal rules about how information can and
11 cannot be used.

12 Having lived through the, kind of, first
13 generation of the interaction between the internet and the
14 web and lots of legal issues like privacy and freedom of
15 expression, and security, there's a tendency sometimes to
16 think that one can wave a technical wand over any given
17 problem and have a result. That is, decidedly, not what
18 we're saying. We think that these tools can help with
19 compliance, but don't avoid the question of what rules one
20 is expected to comply with. And hopefully, this kind of
21 technology will give a view towards the -- the sort of, the
22 structure and nature rules that we actually have to be able

1 to put into place.

2 Let me just say a bit about where we're going
3 from here. As I said, and I'll say again, this is very
4 much proof-of-concept technology, it's not available off
5 the shelf or off our -- it is available off of our website,
6 but it won't get you very far, unless you just like to play
7 around with these things.

8 Our next step is to try to test some of these
9 approaches in more real-world pilot project contexts.
10 We're very interested in the question of how -- how these
11 systems can meet the needs of large enterprises, whether
12 government enterprises, or private sector enterprises, so
13 to that end, we're looking for test cases, we're looking
14 for pilot project partners, and it is that way that we'll
15 understand whether the systems that we've built work
16 efficiently enough for the needs of an enterprise, and
17 probably most importantly, understand whether the policy
18 languages that we've written are actually able to express
19 to the kinds of rules that any given organization needs to
20 be accountable to.

21 So, thanks very much, and I look forward to
22 questions.

1 [Applause.]

2 Mr. Daniel Weitzner: The slides didn't make it
3 into the packet, my fault, but they are on the back table,
4 and there is an article, as well, about our work. If you
5 look at the slides, you'll find some links to some of our
6 papers, and also some of the technology if you feel like
7 playing around with it yourself.

8 So, thanks very much.

9 Mr. Peter Sand: Thank you very much, Danny.

10 Let's just take a few moments to talk about some
11 of the things that have been raised so far, and then go
12 into the last presentation.

13 One of the -- I'll just start off with a
14 question, but I invite the rest of the panelists to ask
15 each other questions, or raise issues that have been --
16 bubbled up from the presentations, so far.

17 One of the comments from a previous panel that
18 really got my attention was that it -- I think it was Steve
19 Dennis from DHS S&T, who talked about rules in a slightly
20 different context. What he basically said is, if you can
21 get two people to agree on something, then that becomes a
22 policy than you can actually implement in technology.

1 And I'm wondering if there are things about the
2 technology that drive the nature of the rules themselves --
3 when we talk about being able to go back and audit
4 compliance with certain rules, or when we talk about
5 building rules into technology, is there something inherent
6 about the structure of technology that requires these rules
7 to be stated in certain ways? Are there limits that you
8 have to place on the rules that can be built into
9 technology, that are driven by the nature of technology
10 itself?

11 Mr. Brian Tretick: I guess I'd like to start out
12 and say the first thing -- from an auditability perspective
13 is, we like the rules to be objective, measurable, complete
14 and relevant. And so, in some cases you might find a
15 situation where we set one rule or two rules when there
16 might be 20 that are relevant to make a complete set of
17 rules, again, just to make sure you're doing all that you
18 should be doing, not just focusing on one thing.

19 The other piece of that is, the measurable and
20 objective piece. But the measurability is important,
21 because you could have a rule that requires you to -- you
22 can't prove something, again, that's immeasurable, and I'll

1 try to come up with an example as we keep going, because
2 one won't come to mind right now, sir.

3 Mr. Daniel Weitzner: That's a good question -- I
4 guess I'd say, overall, rules that can be expressed as
5 specific prohibitions are a whole lot easier to instrument
6 than rules that reflect, kind of, subjective judgment
7 calls. So, a rule that says -- as you had in the example,
8 that K. showed -- you can not use health information to
9 deny service, is a rule that we can describe in a way that
10 a computer can do something with it, and where we can
11 detect a violation, because we understand what health
12 information is, and we understand what a service denial is.

13 A rule that says -- that has reference, for
14 example, to reasonable expectation of privacy, is one that
15 our system is just going to kind of say, "Well, you're on
16 your own on that one." And that's not necessarily a
17 disaster, but it -- so some rules may kind of end with that
18 question, but you know, specific prohibitions are better.

19 The second point I would make is that a lot of
20 the time that we've spent on this work has actually not
21 been on designing new algorithms or spending lots of time
22 programming, it's time spent with people like K., and

1 engineers, who look at how to describe an environment --
2 both the information in an environment and the rules, in a
3 kind of a consistent framework.

4 So, I think that going forward, and the technical
5 term for that is ontology development, but we don't have to
6 go into that -- what I would say is that, I think that to
7 the extent that there is a commitment to be able to have
8 this kind of accountability, it's going to be important to
9 have an effort to have a consistent view of -- a consistent
10 description of the information in the environment that
11 you're trying to control, and make sure that that's built
12 on.

13 If we have 17 definitions, for example, of what
14 an agency is, we're going to have a hard time understanding
15 how to do Privacy Act compliance of FOIA compliance, or
16 name your legal area.

17 So, a view towards gradually building up a set of
18 consistent categorizations for information and consistent
19 styles of expressing rules, I think, is going to be
20 important.

21 STATEMENT OF MICHAEL AISENBERG, PRESIDENT,
22 INTERNATIONAL SYSTEMS SECURITY ENGINEERING ASSOCIATION

1 Mr. Michael A. Aisenberg: Peter, your question,
2 for me, really previews the opening thought that I was
3 going to use to tee up my remarks which, I think is, in
4 part, reflected in this slide, with apologies to our friend
5 and colleague, Tony Rikowski, is a map of the standards
6 bodies which, as of January of 2007, were engaged in
7 projects addressing identity management that incorporated
8 information security, information assurance and privacy
9 issues, and that number -- the individual ovals, to find --
10 individual committees that had those projects at that time,
11 and actually a more recent update of that slide is in
12 preparation now, with roughly twice the number of standards
13 bodies, going from roughly 50 to roughly 100 now, that are
14 addressing that issue.

15 Those are places where the rules that Danny has
16 just talked about are attempting to be developed by
17 communities of experts, or those who describe themselves as
18 experts, and they will promulgate them for applicability
19 and contexts, ranging from government agencies around the
20 planet, to individual users, in particular market
21 stovepipes -- whether it's banking and finance and
22 securities, or healthcare, or other particular areas of

1 application.

2 So, the question that I think is begged by Peter's,
3 question to the panel is, which rules are you talking
4 about? We are at the ever-widening end of a cornucopia of
5 continuing rule creation.

6 Let me just read from a document that has a date
7 of May 7, 2008, it's from the Office of the Press
8 Secretary, Crawford, Texas, "Subject: Designation and
9 sharing of controlled unclassified information,"
10 interesting acronym, CUI. Paragraph two, "The global
11 nature of the threats facing the United States requires
12 that, a) our nation's entire network of defenders be able
13 to share information more rapidly, so that those who must
14 act have the information they need, and, b) the United
15 States government protects sensitive information,
16 information privacy, and other legal rights of Americans."
17 Definitions, 3(a), "Controlled unclassified information is
18 a categorical designation that refers to unclassified
19 information that does not meet the standards for National
20 Security Classification, under EO-12958, as amended, but
21 is, 1) pertinent to the national interests of the U.S., 2)
22 under law or policy, requires protection from unauthorized

1 disclosure, special handling safeguards or prescribed
2 limits on exchange or dissemination." "(f), enhanced
3 safeguarding, is a handling requirement that means those
4 information so designated is subject to measures more
5 stringent than those normally required, because inadvertent
6 or unauthorized disclosure would create a risk of
7 substantial harm, this requirement is indicated by the
8 marking 'controlled enhanced.'

9 Section headed, "Policy, the CUI Framework,"
10 paragraph 7, "All CUI shall be, a) categorized into one of
11 three combinations of safeguarding procedures and
12 dissemination controls, and be so indicated through the use
13 of the following corresponding markings." And I won't read
14 the paragraphs under them, but the three are controlled,
15 with standard dissemination, controlled with specified
16 dissemination, or controlled-enhanced with specified
17 dissemination.

18 [Laughter.]

19 Mr. Michael A. Aisenberg: That's a press release
20 describing an Executive Order, that has not been passed by
21 the Congress of the United States, it does not impact the
22 classification system that is limited to agencies that have

1 classified data origination authority, which are, actually,
2 very few, but it does intend, by its other nine pages, to
3 apply to every agency.

4 So, the question really being begged for me, is
5 what do agency heads, and organization managers, need to be
6 thinking about in determining what controls to employ and
7 deploy? What will be the rules that will be audited? The
8 universe of candidates for those rules is not simple, it's
9 not precise, it's not easily understood, it's growing day
10 by day, and the challenge of picking the rules, and
11 understanding what those rules mean, has to be addressed
12 even before reaching a question of auditing against the
13 rule that you've picked, or in Danny's architecture, being
14 able to test whether there's some validity to the
15 deployment of the rules, and the auditability.

16 So, I think we have a series of problems that are
17 -- it's almost monumental -- I won't say insurmountable, but
18 certainly monumental -- to reach, in terms of the selection
19 of, what are we going to do to protect individual,
20 personally identifiable information both in the government
21 context and in the general marketplace, and what is going
22 to be mined?

1 Ms. K. Waterman: I want to take the opportunity
2 to respond to that, because I think in a way, it actually
3 goes directly to what Danny was saying about driving
4 towards consistency. The CUI project that Michael's
5 referring to came about because it was discovered that
6 there were many different labels people were putting on
7 things that they wanted protected, but that wasn't
8 classified, that could be private -- privacy, which is not
9 part of the classification system. It could be things that
10 were going to be bid out for contract before they became
11 public knowledge, it could be employee -- well, that would
12 still be privacy promotion records, things like that, but
13 there were a large number of categories of things that had
14 nothing to do with classification, and they had labels.
15 And every agency made up their own label.

16 So, some of them used "Sensitive but
17 Unclassified," some of them used, "For Official Use Only,"
18 and it was ultimately determined that there were over 400
19 of these. And as we began sharing with State and local
20 agencies, they complained bitterly that they were getting
21 too many labels, they didn't know what to do with them, and
22 could somebody, please, make this more understandable.

1 So, exactly what Danny was talking about in the
2 consistency standard -- while I agree that, perhaps, the
3 titles are a mouthful -- what's been done is take in over
4 400 labels, and reduce them to three, so that those people
5 getting them could understand they had one of three
6 obligations for how to handle what was being passed to
7 them.

8 So, that kind of consistency, if it actually gets
9 implemented, would be fabulously helpful to the ability to
10 do something at the technical level.

11 Mr. Peter Sand: Another issue which came up in
12 the previous panels, I'll categorize it as a scope issue,
13 which is -- there was a general framework that was
14 laid out in terms of data mining that was basically the
15 data collection, the analysis, and then some kind of
16 decision making at the end. And as you looked at auditing
17 complex systems like data mining systems, are you looking,
18 or would you look just at the internal operation of the
19 system itself, or would you look at a broader context, in
20 terms of what comes out of it?

21 One of the issues that we talked about a little
22 bit in the previous panel was kind of the output of a data

1 mining process. Does it deliver a specific answer? Yes,
2 no? That's the guy? Or is it delivering a probability,
3 like, well, maybe this category of things is interesting
4 and you should look at it.

5 And I would think that as somebody's looking kind
6 of from a distance at one of these systems and where it
7 fits in an organization, the issue would be what message is
8 coming out of the data mining system for the person that
9 has to make the decision.

10 In terms of your reviews or as you look at these
11 instructionally, do you look at a larger scope, do you look
12 at where this system fits in a broader context? Or, how do
13 you decide what's within the scope of your audit, and
14 what's left for the people to work out?

15 Mr. Brian Tretick: I'd like to take a first cut
16 at that, one, you generally audit something that happens,
17 so it's more than just the technology, but it's the entire
18 business process around that. So, from a financial
19 auditing perspective, it's the -- you audit the integrity
20 of the financial reporting process. For,
21 under Sarbanes-Oxley Act, section 404, part of this
22 accountability, you audit the effectiveness of internal

1 controls over financial reporting.

2 In a situation like this, we would look to the
3 rules that are applied, try to map them to some framework,
4 generally accepted privacy principles is one of those, and
5 we would look, then, at the technical, procedural, physical
6 controls, and even the legal controls that are in place to
7 determine whether they were designed in place, designed
8 effectively, and operating -- designed and adequately
9 operating effectively over time -- to determine whether or
10 not we could develop, again, a position of
11 reasonable assurance that there was a controlled environment
12 in place.

13 That way, if you have that kind of controlled
14 environment, you don't need to do the other thing we call,
15 the other end of the spectrum, which is substantive
16 testing, where you use Danny's tool, K.'s tool for every
17 transaction that occurred to see whether it was good, bad,
18 or ugly. One would, in a financial audit, if there were no
19 technical controls, one would have to hull some huge,
20 statistically relevant number of transactions to develop
21 the position of reasonable assurance, and not absolute
22 assurance, that you'd look at all of the transactions.

1 If I could trust the process from a physical,
2 procedural, technical and legal perspective, I can rely --
3 and I can see that it was designed, in place, designed
4 appropriately, and operating effectively over time, I
5 wouldn't have to do that kind of substantial audit, and I
6 wouldn't even have to know what went on, because I was in a
7 trustworthy environment as it is.

8 So, you really do need to look at everything.

9 The short answer is yes, I guess.

10 [Laughter.]

11 Mr. Daniel Weitzner: So, part of what our -- I
12 think this is a -- it's a very useful distinction that
13 Brian makes between the kind of audit that is traditionally
14 done in a financial setting which says, "Is the system such
15 that you can trust that the right thing happened?" I think
16 what we're trying to do is to -- so we're not really trying
17 to build an audit system. We're trying to build a system
18 which helps make sure that the right things happen as they
19 do, and -- as they are happening.

20 So, the analysis that K. showed can be applied
21 either if, for example, that hypothetical customer had a
22 complaint and said, "Well, I was denied service for an

1 illegal reason," or, hopefully to save everyone trouble,
2 the moment that that customer service agent said, "Well,
3 I'm about to -- I'm going to deny this guy service, because
4 I just got a look at his health information." Right then,
5 we ought to be able to tell the user of the system, "You
6 can't do that." Or, at least, "If you do that, you will
7 have broken some law."

8 So, we're really about visibility into the
9 information flow as it happens, and then providing the
10 ability to go back and check, whether either for audit
11 purposes or for other forensic purposes.

12 But, I think what that takes, to come back to
13 Pete's question, it really does take maintaining, a sense of
14 the context of the information. People in
15 computer science tend to call that provenance, where did
16 the data come from? We're interested in kind of a policy
17 provenance, what are the rules that come with the data and
18 how do we keep track of that? So that the person who uses
19 the results of some data mining activity, and then takes
20 that somewhere else and communicates it to somewhere else,
21 is able to easily maintain the context from which it arose,
22 and whatever restrictions would have applied.

1 Mr. Michael A. Aisenberg: I think what's
2 interesting about this question is that it begs the
3 interest of whose interest is being vindicated in the
4 process. Typically, the reason an audit is imposed is
5 because the institution is seeking validation of its own
6 behavior in some community. You're trying to assure that
7 the managers of an organization have followed some rule
8 that they're supposed to follow. And it's their own
9 interests that's typically being vindicated.

10 What Danny's tool adds as an interesting aspect
11 to the audit and conformity community's toolkit, is now you
12 have an additional tool that, perhaps, is available to
13 third parties, like the subjects of the PII, to determine
14 whether an appropriate or inappropriate use has been made
15 of the data -- it expands the universe of remedies in the
16 community of those who are beneficiaries of remedies in the
17 data life cycle community. Now, the PII subject has a
18 potential tool that can demonstrate that there has been an
19 abuse, or a non-conformity with the rule as to how that
20 data is supposed to be handled.

21 To me that's an interesting -- a very interesting
22 development, but it still begs the further question that I

1 asked, or addressed in response to the previous question
2 which is, which rule? And, to me, that still is a much
3 more difficult question in today's environment, because we
4 are -- in this era of massive rule transformation -- coming
5 from an enormous number of sources, from industry-developed
6 internal best practices, even from individual company
7 practices, and industry practices, sector-wide practices,
8 standards bodies, individual nation-states, global
9 standards bodies -- the sources of rules are enormous. And
10 even in the face of a statute, looking for conformity
11 across 86 major U.S. government agencies and the way they
12 will apply those rules is going to be, you know, just
13 complying with the CUI standards -- I guarantee you, you
14 will have a wide diversity of deployment across the
15 agencies as to how they select which aspects of the CUI
16 memorandum they are going to emphasize, between DoD and the
17 intelligence agents. He's at one end of the spectrum, and
18 the more human-oriented agencies like HHS at the other end
19 of the spectrum.

20 Ms. K. Waterman: I agree with everything that
21 Michael has to say here, and I just wanted to add something
22 that didn't come up in our demo. In trying to be flexible

1 with this project, you can apply a rule at any point. So,
2 you can have the system deploying a rule set, which is what
3 people who own the data think is involved, but if, at a
4 later point, you want to audit because somebody's raised
5 the question of a different one, you can actually deploy a
6 different rule.

7 Mr. Peter Sand: Thank you very much.

8 Let's go into our last presentation, and then
9 hopefully we'll have some time for some more panel
10 discussion, and then open up for your questions.

11 Michael?

12 Mr. Michael A. Aisenberg: Thank you, Peter.

13 I had not intended to work from the slides, but
14 as the discussion has gone on, because I would have the
15 propensity to wander and fill most of the rest of the day
16 with observations, I'll use the slides to just to try and
17 impose some self-discipline, here.

18 And the reason I had wanted to lead off with this
19 slide about the standards community was to make a point
20 we've just been talking about -- that the sources of rules
21 today in this space are enormous. And we are hardly at a
22 punctuation point, where any definitive understanding can

1 be asserted. There's nothing to lock in. And even now as
2 we approach a national Presidential transition point, the
3 likelihood of rule and policy changes in the area of data
4 management and data safeguarding privacy across the United
5 States Federal government is likely to be upon us for the
6 next year or 18 months, irrespective of party. It's
7 partially a cultural reality that the new guys like to do a
8 little housecleaning, both at the personnel level and at
9 the policy and practice level, and I think we can
10 anticipate policy and practices regarding data and data
11 collection, and data storage, and data management, and data
12 security to be upon us for the next 18 months, and it won't
13 come just from statute and regulation, it will come from
14 individual agency behavior, individual manager behavior,
15 and in the technology space, it will also come from these
16 emerging standards bodies.

17 I guess my point here is that context is
18 important. And data mining and data exfiltration are
19 perhaps understood as two faces of the same cube, that what
20 may stand for a perfectly permissible practice and behavior
21 by one entity may, in fact, be viewed as something akin to
22 an act of war through a different lens.

1 We have to remember that the networks that we're
2 talking about where these pieces of data are stored are
3 global, and that the operators of the networks are not
4 necessarily U.S. citizens, or U.S. actors are not
5 necessarily acting exclusively in the United
6 States, they are acting globally, and the threats to those
7 data sets and networks are global.

8 So, policies that we make, and policies that we
9 follow, and policies that we observe being followed
10 elsewhere, have to be understood in that global context.
11 And rules that we write may well, as consumed us in the
12 late nineties with the years of negotiation with the E.U.
13 over their privacy principles, may require extensive, and
14 perhaps even unsatisfying -- agreeing to disagree -- kind
15 of outcomes. I think it is important that we understand
16 this.

17 A further contextual hand grenade -- someone
18 simply described it benignly as a landmark, I think it's
19 more of a landmine -- is the Administration's cyber
20 initiative. Finally, after 6 years, there was a public
21 recognition that massive amounts of classified data are
22 being exfiltrated by our adversaries, and the Director of

1 National Intelligence has, since October, been engaged in a
2 program with an investment tag, estimated somewhere between
3 six and \$30 billion to do a number of things. To fund the
4 R&D necessary to make our networks more secure, to enhance
5 the relationship between government network managers and
6 their private sector stewards and contractors, and to take
7 a hard look at the use of personally identifiable
8 information that exists across networks, and understand
9 what rules are being applied by agencies, including defense
10 and intelligence agencies, may act as either enhancements
11 or deterrents to that condition of exfiltration and
12 sensitivity and vulnerability that the initiative is
13 designed to stop. So, I think that initiative will
14 certainly transcend the election and into the next
15 Administration, and it's a major artifact in the landscape
16 of data security and data privacy. And I think it bears
17 watching.

18 Another point that I would suggest is that we
19 should not become too enamored with technology as the
20 exclusive solution. I think you've heard both Brian, and
21 Danny, now describe the importance of the other two legs of
22 the triad -- it's people, processes and technology, and I

1 think we've under emphasized the people and processes part
2 of the triad, in favor of the technology. The tools are,
3 essentially, in most cases, neutral. One person's tool can
4 become another person's weapon, depending on the intent and
5 how it's used.

6 It's very important that we not overstate the
7 capacity of any particular tool to do the job. Right now,
8 cryptography is considered to be the tool of choice for
9 protection, and it appears, specifically, in all kinds of
10 statutes and regulations. However, a recent paper
11 published out of Princeton, and you see the URL for it up
12 there, and the exact title of it, if you will bear with me
13 for a minute, is Less We Remember, Cold Boot Attacks on
14 Encryption Keys. And basically, it's a 12-page paper that
15 describes the human behavior that is necessary in order to
16 make strong encryption truly effective, to recognize the
17 vulnerabilities that exist when data -- even encrypted data
18 -- is stored on a live memory chip, on a live board, in an
19 active processor that happens not to be turned off. The
20 simple benefit of rebooting, and clearing the cache, has an
21 enormous impact on the security of the system, and the
22 ability of that cryptographic system to achieve its

1 intended benefit. And those who fail to turn the system
2 off suffer the penalty of opening up certain means of
3 access, of that key to those who have the technical tools
4 to decipher and attack the data in storage.

5 The point being not that crypto isn't a good
6 tool, but that every good tool has to be deployed in an
7 effective manner, and this year's good tool may be defeated
8 by next year's better tool cutter. One of the first
9 lessons I learned in security was that you go back to basic
10 principles from the 12th Century -- for every chain that
11 you deploy in the castle keep to lock up the prisoners,
12 there's going to come along a strong chain cutter to cut
13 that chain. And so it's a constant effort of mutually
14 assured destruction, if you will, nuclear arms race kind of
15 environment of constant pursuit of the encryption or other
16 protective tools, and the constant battle by the bad guys -
17 - and in some cases the good guys -- to develop the tools
18 to defeat the last generation of protection with the next
19 generation of cracking. And unless we are aware of and
20 sensitive to that constant evolution in the marketplace, we
21 run the risk of writing particular technologies into
22 practice, into regulation, into statute, in a manner that

1 leaves us very vulnerable.

2 Flexibility, and the ability to deploy the most
3 appropriate approaches to protections -- and that includes
4 not just the technology, but good practice, and good
5 inquiring about the individuals who are the practitioners.

6 One of the major areas of vulnerability in the
7 Administration's Cyber Security Initiative, is taking a
8 real hard look at the people with the highest levels of
9 clearances, in the highest levels of our own national
10 security and cyber-security community. Because the insider
11 threat -- or, as some in the industry like to describe it,
12 the threat that exists in the 18 inches between the screen
13 and the back of the chair -- is often overlooked, and very
14 often where the worst offense is occurring.

15 One last point, the last quote on this page, and
16 this is an old homily, but built-in is typically cheaper
17 and often easier than bolted-on. As you design system,
18 baking the security, and baking the privacy protection in,
19 is typically a better approach than waiting until the
20 system has been developed, and then coming back after the
21 fact to bolt it on. That's a historical lesson that goes
22 back -- for the engineering and technical wonks in the

1 room, there was lots of good security in the X-500
2 architecture, but when we developed X-509 TCPIP, the basic
3 framework that now modern desktop computing is based on
4 today, an awful lot of the security was left off because of
5 the scalability issues, now we're facing the consequences
6 of that by trying to bolt on security to the TCPIP
7 environment, at a cost of billions of dollars that,
8 perhaps, could have been avoided if we had thought about
9 this or recognized the risk 20 years ago.

10 Finally, I would suggest that the path forward
11 ought to be one -- and I think this is sort of implicit in
12 a lot of the work that Danny and the folks at MIT have done
13 -- there's a reason why PGP is such a fascinating name.
14 The P doesn't stand for Perfect Privacy, it stands for
15 Pretty Good Privacy. Eighty percent solutions may be not
16 only the best we can do, but more than enough to overcome
17 the threats that we're facing, to deter the bad guys, to
18 send them off looking for other avenues of achieving
19 whatever their ends are. And the relentless pursuit of
20 perfection may be an interesting advertising slogan, and
21 appropriate to auto engines, but it may not be necessary in
22 this environment.

1 What is more appropriate, I would argue, is the
2 relentless pursuit of progress, that we should be trying to
3 do better than we have been in the past. That includes
4 policies that don't focus on locking the barn door after
5 the horses have escaped, which is what I think breach
6 notification is all about, but end-to-end data custody
7 practices that instill both in the subjects of the PII, and
8 all of those who are depository institutions, whether they
9 are those who are able to compel deposit, like government
10 agencies, or obligatory ones like universities, or
11 voluntary ones, the people that you transact, give up
12 credit card data with, as you're doing online transactions,
13 whether it be eBay or whoever.

14 They have varying levels of obligation, I would
15 contend, to protect that data when they hold it. The
16 greatest burden, I would argue, falls on government who --
17 because they can compel the deposit of data, and the
18 gratuitous third-party folks, the data brokers with whom
19 individuals have no relationship, no legal privity, who
20 perhaps don't even know that those parties hold their PII,
21 and yet those parties are making money off of the use.

22 When it's abused, those folks, I would contend,

1 ought to be pilloried and given the worst that our legal
2 remedy system allows us to impose, but for my money, it's
3 insufficient for us to develop those structures
4 unilaterally as a nation-state. We have to do that in
5 collaboration with our allies, and those who may be not be
6 -- necessarily be our allies. These systems, in the face
7 of the global networks I described at the outset, have to
8 be addressed on a global basis.

9 That's a challenge. Getting the Chinese to come
10 to a table for a diplomatic discussion about common
11 remedies for privacy abuses is not going to be an easy
12 thing to do, but neither is discussion of global warming or
13 nuclear nonproliferation -- the important tasks facing us
14 as a global community going forward, and this is sort of a
15 transition, I guess, to the next panel, and the discussion
16 of policy.

17 But the transition to that has to be, in my view,
18 done on a global basis, both with our close friends and
19 those who are not our friends, and we have to be willing to
20 invest in a long-term negotiation, in order to achieve the
21 benefits that are commensurate with the enormous economic
22 and national security value that these technologies now

1 provide for us.

2 My friend, Steve Malfrus, at the Fed, likes to
3 talk about -- we're a \$12 trillion, going to \$13 trillion
4 economy, well guess what? \$3.5 trillion of economic
5 activity happens over the network every day. Most of that
6 is initially handled on secure EDI networks that are
7 Electronic Data Interchange networks that are managed by
8 the Federal banking regulators, but without
9 bankofamerica.com and NYSE.net and treasury.gov and
10 barkleys.uk -- without a secure, safe, and available
11 internet, to receive that data -- never mind nuclear
12 command and control codes and the national security stuff -
13 - the fundamental economic juice of the United States and
14 the Western economy will begin to grind to a halt at the
15 rate of \$135 billion an hour. If that's not a number that
16 catches the attention of CEOs and Congressmen and agency
17 heads, I don't know what number will.

18 So, there's a lot at stake here, we have a lot of
19 work left to do. But as you can tell from the discussion
20 over the last day and a half, an awful lot of important
21 work has been done -- now it remains for the policy
22 community to make choices from among the candidates for

1 rules, to figure out what the right rules are to deploy,
2 and help people understand how to use them in an effective
3 way.

4 Thanks.

5 [Applause.]

6 Mr. Peter Sand: Thank you very much, Michael.

7 Again, going back to one of the themes that was raised in
8 an earlier panel, which I think your talk really pressed on
9 here is, whether data mining systems or technology itself
10 present unique challenges, or one of the earlier panelists
11 talked about data mining tools as more, like a power tool.
12 It does what you've been doing, but it does it better, it
13 does it faster, it's easier for you, it's more efficient,
14 more effective. And I'm wondering if there, if that's what
15 we're looking at when we're talking about data mining, is
16 it just doing the traditional things, but better, faster,
17 stronger, cheaper, or is there something qualitatively new
18 about using technology to do these same kinds of operations
19 and rules-based behaviors that raise new challenges that we
20 need to address separately from the underlying processes
21 that would happen without the computers.

22 Mr. Daniel Weitzner: I think that what's

1 qualitatively new about data mining, which I gather
2 yesterday you decided you were not going to define. But, I
3 think what's qualitatively new is the scale. It's the scale
4 of information that's available and the analytic power over
5 that information, the social network example I gave, it's
6 just one kind of little toy indication of that relative to
7 what I suspect is available in places that the public can't
8 see.

9 And I think that, you know, to Michael's point,
10 when the problem that we had was securing, you know, a
11 small number of very high-value assets, like nuclear launch
12 codes, or even when the problem was securing relatively
13 well-defined networks amongst a defined number of financial
14 institutions, we could take one approach, and I think that
15 approach in many cases was characterized by, you know,
16 very, very innovative use of cryptographic techniques to
17 keep information out of the hands of people who shouldn't
18 have it.

19 I think that what we're learning, even in network
20 security, but I would say even more so in the kind of
21 information accountability that we now need is that, as
22 Michael said, to close the barn door after the horse is out

1 is not going to be a satisfying result any more.

2 So, I think that as we get to a kind of web-scale
3 analysis of information, and by web-scale I just mean big,
4 a lot of it, I think our ability to control, up front,
5 every single possible use of that information, is going to
6 decline, so what we have to do is supplement the
7 traditional security techniques -- we don't throw the
8 traditional security techniques out, we don't just say,
9 well, everything's now free, and public and available.
10 But, I think what we have to recognize that there are
11 limits to our ability to control information flow and
12 information use up front, and we have to supplement them
13 with the kind of auditing and analysis techniques that
14 detect behavior or harm that we want to prevent against.
15 And this is where the 80 percent rule is going to become
16 awfully important.

17 Most of the rules that we live with in the world,
18 particularly in the financial world, most of the rules are
19 not enforced up front. Most of them are enforced by virtue
20 of the fact that people know they will get caught if they
21 break them -- the tens of thousands of publicly traded
22 companies that have to tell the Securities and Exchange

1 Commission what they're doing, file huge amounts of
2 information with those regulatory agencies, and no one in
3 those agencies look at that data, no one checks whether
4 it's correct or not.

5 Other people may check -- do check -- when
6 there's some reason to look, and then the penalty for
7 misreporting, or false reporting, is very high, if you get
8 caught, but mostly people comply with those rules because
9 they believe they would get caught and there's a, sort of a
10 credible threat along the lines of, you know, sort of a
11 credible, trusted system that you could audit, that you
12 could detect and audit, that will catch the violations.
13 And I think we have to stop treating personal information
14 as this sort of special secret commodity, it is special,
15 but it's just less and less secret. And we have to, I
16 think, start thinking more about the kind of regulatory
17 models that I think are largely successful in very large-
18 scale institutions, like the financial markets, that
19 recognize widespread flow, but still have expectations
20 about ultimate compliance.

21 Mr. Michael A. Aisenberg: Two quick comments on
22 what you just said, Danny, one on the issue of scale, lest

1 anyone think that it's not getting huge -- the daily number
2 of resolutions on the network that -- resolutions means
3 either an email address being sought to direct a piece of
4 mail, or a URL that you type into a browser, trying to
5 direct you to a website, has been hovering in the 500
6 billion resolutions per day number for the last 12 to 18
7 months -- 500 billion resolutions per day.

8 It grew from about 250 billion at the start of
9 this decade, but it has plateaued. Some of that is
10 capacity related, but as we go from a billion users outside
11 of North America, to 2 billion outside of North America,
12 it's expected that that scale will rise, and on the SOX
13 point, in terms of cost.

14 I think it's public information that when I was
15 at Verisign, the first year that we had a SOX-compliance
16 audit, Stratton Sclavos was CEO, complained bitterly in a
17 public forum that it had cost him an additional \$16 million
18 to pay his auditors for that SOX conformity. And less
19 than a year later, an HR employee's laptop was stolen,
20 which resulted in a roughly \$5 million cost to pay all the
21 dismissed employees whose PII was on that laptop, to get
22 them all three credit report free-year-of-service remedies -

1 - that seems to be the prevailing way that you deal with
2 stolen PII from laptops from employers these days.

3 So, you can spend a lot of money on Sarbanes-
4 Oxley conformity, and still not necessarily get a cast-iron
5 defense against bad things happening.

6 Mr. Brian Tretick: Excellent.

7 I've got a couple of points. First, to go back
8 and say that, again, I would contend that the legacy of the
9 financial audit analog works here with data mining, works
10 with whatever you're doing, on privacy. And as we explain
11 it, the financial audit is developing assurance over the
12 collection, use, storage and retention of financial
13 information -- disclosure of financial information --
14 integrity of the process. There's confidentiality
15 requirements before public release, so everything is in
16 there.

17 It's the same deal with personal information, or
18 other confidential information -- trade secrets or
19 government-classified data - have controls over the collection,
20 use, retention, and disclosure of that data.

21 The AICPA generally accepted privacy principles
22 can be seen -- not just as a testable set of standards, and

1 auditable standards -- but also can be seen as an API, like
2 an Application Program Interface, where we have the myriad
3 of requirements that are at the agency level, or the
4 department level, or the business level, or the legal or
5 regulatory level. We believe these can be mapped to this so
6 that

7 you can build programs of controls and assurance that see
8 all those myriad of requirements through the lens of
9 generally accepted privacy principles.

10 Like a medical examination, a privacy audit may
11 not always just tell you you're in good health. It could
12 tell you that you've got to drink less, eat less fatty
13 foods, or that you're near death.

14 So, in some cases that'll tell you -- a privacy
15 audit will tell you that a control isn't in place; it isn't
16 designed adequately. It's weak or it can't be shown to be
17 operating effectively over time, or in some cases, it's
18 just not auditable; it's just not knowable.

19 What this does -- these qualified opinions,
20 everything's good by these 6 things - these qualified
21 opinions point to risk, and would allow management of the
22 organization, or the partners in a data-sharing
23 environment, to point to somebody and say, "You need to fix

1 that."

2 Okay, so that's the two -- so, the -- and the
3 last point doesn't quite fit neatly with all of these
4 others. We look at three types of controls: we look at the
5 prevent controls -- stop bad things from happening,
6 encourage good things to happen, okay? That's the first
7 level.

8 The second is the detect controls -- if something
9 bad does happen, or starts to happen, or whatever, tell the
10 right person so they can be reacted to.

11 And the third are things that correct the
12 situation. And your rules may not be prevent, prevent,
13 prevent; it could be, here's a little prevention, a little
14 detection, and a little correction, and that may be your
15 rule set, and you can operate it imperfectly -- if you have a
16 forest fire, again, preventing forest fires is a good
17 thing. But, if you have a forest fire, you have a process
18 to contain it. And, in many ways, that acknowledgement
19 that there could still be a forest fire, that you can't
20 prevent them all, is probably the most healthy way of going
21 about your business.

22 Mr. Michael A. Aisenberg: I think what Brian has

1 just described is what Toby certainly will remember from
2 the FTC workshops on privacy and security, the homily
3 that's been around for some time, that you can have
4 security without privacy, but you can't have privacy
5 without security. Brian's three points are the essence of
6 systems security deterring, preventing and defending
7 against incursions. You do your best to keep them from
8 hitting you when they are there; you try and make them
9 bounce off of your defensive tools, and when they finally
10 make it through, you try to mitigate and minimize the
11 amount of damage they do to your systems. That's the
12 nature of the security, and without those security tools in
13 places, any sort of privacy architecture is going to be
14 incapable of being deployed and achieved effectively.

15 Mr. Peter Sand: Well, thank you very much for
16 the presentations, it's been really helpful, thank you.

17 [Applause.]

18 Mr. Peter Sand: And I'd like to open up the
19 floor to your questions, if you have any, please come to
20 the microphone and ask us what you like.

21 Mr. Bob Burns: Good morning, Bob Burns, HSR, DHS
22 Science and Technology Director. I really enjoyed the

1 presentation on your system. It was amazing. Yesterday we
2 talked about the complexity of data, showing all its trees,
3 and if you touch this point, you go here; you're showing
4 the converse with the rules and the regulations and the
5 impact to the world we live in.

6 The question has to deal with not the person who
7 was refused service; the question kind of deals with the
8 person who started it all: the patient in the coma.

9 Everybody acknowledges, okay, this person had tuberculosis,
10 and we needed to find out who he had been in contact with -- we
11 did a privacy search, I gather, in your theoretical model,
12 but it would happen in the real world, too.

13 So, in that case, a diagnosis was made which
14 validated us doing data mining, and therefore, taking an
15 action. And everybody says, "good," because the threat to
16 the populace, I guess, exceeded the need for the person's
17 privacy.

18 So, the question is -- you mentioned this earlier
19 in your brief, and it kind of opened it up to everybody --
20 assessing data, using the data, in the data mining process.
21 Would it be reasonable - or does it get too far into the danger
22 Zone - can we use the assessment of the data, the analysis

1 of the data, as the diagnosis phase?

2 Then we have to make the judgment call how to use
3 the data, and is that taking it too close to where you
4 might have mission creep, or the danger zone, so to speak,
5 because what we're finding, or what I see in many cases, is
6 we're not able to make the diagnosis, so we can't decide
7 what we need to do.

8 And in a lot of the cases, the things we want to
9 diagnose either haven't fully occurred yet, such as a TB
10 case, or we're trying to prevent them because of the nature
11 of what would occur. How do we balance that? The
12 assessing, putting in the safeguards, so that I can make
13 the diagnosis and make the decision to use it, or am I
14 getting too close to that danger zone?

15 Ms. K. Waterman: If I understood your question
16 correctly, I think from our -- from the process of a system
17 like ours, at each step where data is moving, or data is
18 being used, you can apply the exact same analysis. So, if
19 the question is, at the very front, can the CDC use the
20 information it has to conduct an investigation, to go out
21 and seek more information? There are rules about that, as
22 well, and you do exactly the same process ; you would take

1 the information it had at that time, and its attempt to go
2 out and collect, and run the rule over to see whether the
3 collection was permissible.

4 Mr. Daniel Weitzner: Could I just -- let me just
5 see if I understand your question. I think you're asking
6 whether it's acceptable -- according to some set of
7 standards, and I'm not sure what the standards are here --
8 but is it acceptable, according to some notion of privacy,
9 to generate what you might call the reasonable suspicion or
10 the probable cause from the data? So, is it okay to look
11 first, in order to find the justification that we would
12 normally expect to continue with an investigation?

13 And I'm going to dodge -- if that's your
14 question, I'm dodging it, and I'll tell you why I'm going
15 to dodge it. Because I think, as you know, the answer is,
16 that it depends. That there are times when, you know, when
17 traditional Fourth Amendment law, when we say there are
18 exigent circumstances, and go back and prove that, you
19 know, prove that you had probable cause later, or get the
20 order later. There are times - we picked the public health
21 example because it's what people tend to come up with as
22 the break-the-glass, sort of, example -- it's an emergency.

1 You don't have time to ask permission of the patient who's
2 in the coma, or the -- whoever else it is. You go; you
3 need to find the answers.

4 Now, I don't -- I guess my belief is that amongst
5 the range of, sort of, analytic activities that are
6 possible, there are going to be times when we say, "Yeah,
7 break the glass, and prove that you were right in doing it
8 later." And there are going to be times when we say, "No,
9 the information is too sensitive, the activity is too
10 intrusive, and we're not" -- and simply the fact that we
11 could find some evidence of illegal activity is not
12 sufficient to say, "Let's just sort of sweep up every piece
13 of information about every person, and see what we find."
14 That's the extreme. And, I'm not --

15 So, I guess what I would say is that we've tried
16 to design a system that allows us to operate across that
17 spectrum, so that if justification is required before doing
18 the analysis, we can represent that. If what you want to
19 be able to do is retroactively establish justification at
20 some later point, we could keep track of what happened, so
21 that the actions that were taken can then be assessed to
22 see if they would have been justifiable.

1 And I -- so, from a technical standpoint, I don't
2 actually think it's our job to build a system that is
3 biased towards one end or the other of that spectrum,
4 because I think all conditions are going to happen. What's
5 important from our standpoint is to make sure that in --
6 wherever you are on that spectrum -- that you can establish
7 either in advance or after the fact that you actually
8 followed the rules that were in place.

9 Mr. Bob Burns: And your dodging -- to use your
10 term -- is fine, because it drives home a point that we
11 came up with yesterday, as well. When trying to ascertain
12 what is acceptable, what we can look for, how we can use
13 the data, I understand completely why you go with the
14 tuberculosis patient example -- everybody rogers it; it's
15 locked in law; you know what you can work with. We're
16 having the problem -- how we come up with what is back to
17 Congress, perhaps, or legislation, or rules, one of the
18 things we can look for, and what are those paradigms that
19 we can work within, like the tuberculosis patient, how do
20 we make that happen? How do we drive that?

21 Ms. K. Waterman: One thing I just want to loop
22 back, and then -- Michael, if he wants to answer -- what

1 Danny talked about before with the dependency tracker is,
2 we have the ability, if somebody's established criteria,
3 which, in most cases they have, for what is probable cause
4 or what is a reasonable belief. If we can define those
5 things, the system can at least pop up what facts it has,
6 and some person could make a judgment about whether that
7 rose to the level of enough or not enough, but you could at
8 least percolate the facts back up.

9 Mr. Michael A. Aisenberg: I think this is an
10 area where we have to be careful again, and I personally
11 believe that the technologies are, in virtually all cases,
12 from a public policy, rule-of-law standpoint, neutral. And
13 the question is, how are the technologies used?

14 And at the end of the day, for me, this goes to
15 the issue that Peter Swire and a number of us
16 were talking about in the back of the room, which is the
17 excellent - I guess it's operating code of the privacy group
18 at DHS - which is Ben Franklin's statement, which is one of
19 my operating rules in this space: "Those who would
20 sacrifice a moment's liberty for the sake of temporary
21 security soon will find themselves having neither."

22 We've heard any number of cabinet secretaries and

1 committee chairs since 9/11 justify what they consider to
2 be compromises and exigent deployments of tools and
3 behaviors and practices in law enforcement and elsewhere
4 over the last 7 years, based on the terrorism threat, the
5 national security compulsion or some other rationalization.
6 And I say, "Wrong."

7 For me, as an attorney, for me as I understand
8 our constitution, for me as someone who is hoping that the
9 world that my grandchildren and great-grandchildren grow up
10 in is not all that different from the one that we've
11 understood, at least in terms of legal remedies, the most
12 important thing is not letting that balancing act become
13 the way we behave in this space and become the excuse for
14 deploying what can become draconian applications of
15 technology.

16 The technology is very seductive, especially when
17 it's poorly understood. We've got two communities: we've
18 got the rule-of-law guys over here, and we've got the Ph.D.
19 computer scientists over here, and there are a couple of
20 folks in the middle who are both, and they try to translate
21 back and forth, but the technology can very quickly get out
22 of the barn and become the devil.

1 And it's the responsibility, I would argue, of
2 those who are part of that rule-of-law community to remind
3 us over and over again of what Franklin said, and make sure
4 that what we try to do is use our genius as Americans to
5 maintain and maximize both liberty and security at the same
6 time, and come up with the tools and with the application
7 of those tools that are consistent with that value set.

8 That's a challenge, but I think we're up to it.

9 Mr. Peter Sand: Thank you very much.

10 Let's move on to the next question.

11 Mr. Chris Clifton: Yeah, I -- almost a follow
12 on, but just to step back, there's -- this panel, I get the
13 feeling that there's -- we're starting to move towards
14 thinking that, or equating privacy with access control.
15 And I think privacy is much more complex, and it's starting
16 to sound like it's a form of access control.

17 As an example, in this scenario we've been
18 discussing, I think there's a very big difference between
19 general use of the information to discover pandemics and
20 this particular scenario, which is saying, "Hey, we're
21 trying to discover people who need our help." If you look
22 at privacy laws, they get it.

1 Use of my data to do something that I -- you
2 know, that is for my benefit, or that I have requested, is
3 very different from using it for some general use. There's
4 also the risk of harm. Using, you know, using my private
5 data for something which may result in my loss of phone
6 service is very different from using my private data for
7 something which may result in my incarceration.

8 And I just want to make sure that we keep in mind
9 that privacy is not about binary -- you see the data or you
10 don't. There's a lot more to it than that. And I'm just
11 wondering, do you have any comments on that?

12 Mr. Brian Tretick: I'll take that as, not to be
13 seen as the mouthpiece of the American Institute of
14 Certified Public Accountants, I am a card-carrying member -
15 -

16 Mr. Daniel Weitzner: Can I see the card?

17 Mr. Brian Tretick: Actually, I don't know where
18 I keep it. Nobody's ever asked me for it before.

19 [Laughter.]

20 Mr. Brian Tretick: We try to define privacy in
21 the -- for the -- what became the generally accepted
22 privacy principles; we ended up describing it, and we

1 described it using these words: privacy is described as
2 the, this kind of tension between the rights of the
3 individual, and the obligations of the organization, over
4 the life cycle of personal information.

5 So, the rights of the individual, obligation to
6 the organization over the collection, use, disclosure, and
7 retention of the personal information, but the use is the
8 big deal. The way I describe it is, privacy is -- it's the
9 wrong word. It doesn't mean anything, but we're stuck with
10 it. Privacy is protect the data and govern its use. So,
11 it's security and govern its use, whatever -- by whatever
12 the rules are.

13 Going back to the first question, as an auditor,
14 I'd look to have control over governing the use. If you have
15 the data and you don't know what the use is, perhaps
16 the governance is some sort of rules of engagement over
17 the data, which could include, for example, an independent
18 review board, or like your FISA court or something like
19 that.

20 So, what we would look for is not necessarily
21 specific controls over the use of the data inherent in a
22 system, but a procedural process and a governance, a data

1 governance overlay, over top of that that would say that,
2 when data is used, it has gone through the right level of a
3 review and approval before it happens. That would be my
4 definition.

5 Mr. Daniel Weitzner: Could I just say, really --
6 Chris, I think that was about the most -- you said, I
7 think, about the most important thing that one could say
8 about privacy, which is that it's more than just access. And,
9 I'm not sure exactly to whom you addressed the question,
10 but I can say that the motivation for our work is to be
11 able to do a better job of detecting uses that are contrary
12 to whatever the rules are. Because I think, as you know,
13 systems are reasonably good at preventing and detecting
14 access, because that's an event that, kind of, computers
15 are better at representing. I think it's harder to
16 represent uses both legally and computationally. So, I can
17 only agree 100 percent.

18 Mr. Peter Sand: Let's move on to the last
19 question.

20 Mr. Dave Weitzel: Dave Weitzel from Mitre.

21 The question's focus is to Danny Weitzner, but I
22 wanted to build on Brian. In the audit world, and as a

1 former Director of PWC, I think I understand a little bit
2 of your business, you have the GAAP standards that go
3 back a long, long time, and they really are generally
4 accepted. And financial controls - we can go through 150
5 years of audit history to see how that stuff's evolved.

6 One of the challenges of GAAP standards is,
7 they're really not generally accepted -- yet, right? We
8 have it by the folks that Michael works with in the
9 security controls industry, so my question to Danny is, in
10 information security space, we have ISO-1799, and its
11 progeny; we have ITL; we have COBIT; we have NIST, FISMA
12 800-Series standards -- how do we make that happen so that
13 people know that we're talking about pick your standard?
14 Maybe we should be starting with the GAAP standards here,
15 but how do we do that for the tools you're building, and
16 build it in a way it can be adopted broadly?

17 Mr. Daniel Weitzner: That's a great question. I
18 think that it has to start, Dave, with an understanding of
19 what the rules are we're trying to enforce and have
20 accountability against. So, I guess I'd poke a little bit
21 at the definition that Brian offered of privacy as this
22 balance between institutional and individual interest --

1 well, that's true, obviously. I'm not sure how much it
2 helps us.

3 It certainly doesn't help me, at all, in thinking
4 about how to design a system; it just tells me that there's
5 a problem, and I think that's -- and I don't think that's
6 the fault of the auditors or accountants. I, frankly,
7 think it's the fault of policymakers who have to say what's
8 allowed and what's not.

9 We, you know, whether you look at the various TSA
10 Passenger Risk Assessment programs, and, in all candor, I
11 see the debate over the last, I don't know, 5 or more years
12 of those kind of programs, as a failure to come to
13 consensus about what the underlying substantive rules ought
14 to be. Once we know those, then there are a whole bunch of
15 procedural rules we obviously have to put in place to make
16 sure that we can enforce those, but I think we've got it a
17 little backwards, frankly.

18 I think we know a lot about good privacy process,
19 and I think that is, in many ways, inspired by the
20 experience of -- well, certainly the experience of fair
21 information practices, and the experience of applying, sort
22 of, auditing and accounting techniques, but I just don't

1 think we have enough clarity about what the actual rules
2 are. And I can say that, in some part, because we've kind
3 of gone around and tried to find scenarios that we can
4 model, what we find over and over again is it's really
5 not clear what you could do with a lot of this data, one
6 way or the other.

7 And I think that the problem is that, a) we can't
8 build systems, and b) I think citizens who, to follow
9 Michael's point, I think have -- really have a right to
10 know what's going on with their personal information,
11 especially as to the government -- simply don't know. We
12 just don't know. And that's because we don't know what's
13 on which side of the line.

14 Mr. Michael A. Aisenberg: I want to vigorously
15 disagree with part of what Danny just said, and vigorously
16 agree with another part of it.

17 The disagreement is with the notion that what we
18 -- and maybe you didn't mean to say it this way, maybe we
19 actually agree across the board -- I don't think the issue
20 is selecting which rules we want to follow because of the
21 rule's inherent importance. I think the issue is, what's
22 the outcome that we ultimately seek?

1 What's the behavior that we're seeking to have?

2 And the condition that we're seeking, in terms of -- you
3 know, we can talk about macro conditions -- we want to
4 relieve ourselves, as a nation, from the threat of
5 terrorism? Or, we want to see that all personally
6 identifiable information is only used in accordance with
7 its originator's intended uses and never abused.

8 I think you have to define certain endpoint
9 conditions before you can select which rules sets in
10 operation will get you there. I think that's one of the
11 great deficiencies that the industry that I've spent 30
12 years in has. The computer and technology industry
13 constantly wants to have satisfied and happy customers, and
14 so it doesn't much care what the customer's end use is.
15 They just want to make sure that the tools they have to
16 offer will meet the procuring agent's demands and the
17 specifications for whatever that particular customer wants
18 to use the tool for, and they don't care what the output of
19 that process is.

20 We've got to start caring, as a community,
21 about the output. We've got to start caring about what it
22 is that we're developing that may be much more suitable for

1 enabling abuses than for enabling protection.

2 Mr. Peter Sand: I'd like to end with one last
3 thing, and that is, if there's one last thing you could tell
4 a developer or a program manager going into data
5 mining, specifically, if there's one thing you could tell
6 them, based on your experience that they should do, before
7 they start or in the beginning, what would it be? Just a
8 quick one-liner, thanks.

9 Mr. Michael A. Aisenberg: Well, I'll just follow
10 on what I just said -- have a clear idea of what your
11 actual intended purpose of your program or project is
12 before you start designing it. If you don't know what you
13 want to do with this technology and this data, then you
14 ought not to be building a system in the first place.

15 Ms. K. Waterman: Going back to something that
16 Michael talked about at the very beginning, I think that
17 one of the really important things to tell a system
18 designer is to make sure that the system has enough
19 available information about what took place or what is
20 taking place.

21 Mr. Daniel Weitzner: I think -- I'm just
22 building on K.'s point -- that the transparency of

1 system operations, both to be able to assess effectiveness,
2 and to be able to measure compliance against whatever rules
3 you feel you're accountable to, are really -- really the
4 important thing. Where it seems to me we're kind of in early
5 days of understanding how well these systems work, what
6 impact they have, both on their national security or law
7 enforcement goals, and on their privacy impact, and we
8 really need as much as possible to have both of those
9 measures available to policymakers in order to chart a
10 course forward.

11 Mr. Brian Tretick: What I'll say is kind of
12 embedded in all of your design -- your systems and the
13 supporting processes, the legal relationships between
14 entities, with the control -- controllable environment and
15 auditable environment -- in mind.

16 Mr. Peter Sand: Well, thank you very much for a
17 great panel, and thank you for participating.

18 [Applause.]

19 Mr. Peter Sand: And we have a break now until
20 10:30, so please be back at 10:30. Thank you.

21 [Recessed 10:22 a.m.]

22 STATEMENT OF TOBY MILGROM LEVIN, SENIOR ADVISOR,

1 DHS PRIVACY OFFICE, MODERATOR

2 Ms. Toby Levin: Welcome back to the last panel
3 of the workshop and a significant -- all of the panels are
4 significant, but I like to think of this panel as the most
5 significant among the significant panels, because it will
6 be our effort to pull together much of what you've been
7 hearing over the last day and a half, and really set us off
8 in what, hopefully will be a very -- give us a roadmap for
9 the direction that will lead our office in the coming
10 weeks.

11 Again, my name is Toby Levin, with the DHS
12 Privacy Office, and I want to introduce you to our panel,
13 many of whom you saw in earlier panels.

14 To my right, Peter Swire, who is the William
15 O'Neill Professor of Law at the College of Law at Ohio
16 State University, Danny Weitzner, who is Co-Director of the
17 Computer Science and Artificial Intelligence Laboratory,
18 CSAIL, Decentralized Information Group, at MIT.

19 To my left, Tom Oscherwitz, who is Vice President
20 of Government Affairs and Chief Privacy Officer for ID
21 Analytics, and then Fred Cate, Distinguished Professor and
22 Director, Center for Applied Cybersecurity Research, at

1 Indiana University, and Barry Steinhardt, Director of the
2 ACLU Program on Technology and Liberty.

3 The end of the last panel really teed up our
4 panel quite well. It ended with the challenge in defining
5 the rules -- what's allowed, and what's not and how do we
6 do that? We're not going to necessarily be able to define
7 the rules, but hopefully we'll be able to discuss a process
8 for moving forward on data mining in a privacy-sensitive
9 manner at the Department of Homeland Security.

10 And we can do that at a 30,000-foot level,
11 20,000, ten, on the ground -- obviously we won't be able to
12 accomplish all of that today, but we will be looking at how
13 government agencies can work with a set of principles --
14 work toward developing a set of principles -- that can
15 guide data mining research.

16 We'll begin with an overview by Professor Fred
17 Cate, on what are the principles that may define best
18 practices based on prior efforts, because we're not
19 starting from a blank slate, and then we'll move on to a
20 panel discussion.

21 So, Fred, we appreciate you starting us off.

22 STATEMENT OF FRED H. CATE, DISTINGUISHED

1 PROFESSOR AND DIRECTOR, CENTER FOR APPLIED CYBERSECURITY
2 RESEARCH, INDIANA UNIVERSITY

3 Mr. Fred Cate: Thank you very much, Toby.

4 And, a special thanks for recognizing that, in
5 fact, this is not the first time the issue of data mining
6 frameworks and legal standards have been discussed in the
7 nation's Capitol.

8 There have been many efforts before, many
9 scholarly efforts, many political efforts, many efforts by
10 industry groups, and others, to try to identify standards
11 for data mining -- standards that would allow data mining
12 to proceed in a way that's useful to protect against
13 terrorism, and to enhance national security, at the same
14 time as to protect privacy.

15 And so, it seems a useful place to start to
16 identify some of the -- what I refer to -- as sort of the
17 low-hanging fruit. These are the places where there has
18 actually been a fair amount of consensus, even if the
19 details have not, in every case, been worked out, of the
20 steps necessary.

21 But, before starting there, it might be useful
22 just to identify three fairly broad conclusions that, I

1 think, virtually all of the groups that have looked at
2 these issues have agreed upon.

3 One is that privacy is important, and we talked
4 about this, at length, yesterday, I'm certainly not going
5 to belabor this now, but that privacy is important for many
6 reasons, and in part, because it is a fundamental condition
7 for most other civil rights to be exercised, as well. And
8 therefore, because of that intrinsic linkage, privacy is
9 worth protecting.

10 Second of all, that data mining does affect
11 privacy -- that, inevitably, when the government collects
12 or uses data about individuals, it affects privacy -- it
13 may affect it to different degrees, depending upon the type
14 of data mining, the consequences will obviously impact
15 that, but that it does so goes without saying, despite the fact
16 that this has certainly been questioned by some leaders in
17 Washington -- that when the government engages in data
18 mining, privacy is affected.

19 And third, that it is possible to both do data
20 mining and protect privacy -- that this is not an
21 insolvable, or unsolvable problem -- rather, this is a case
22 of doing the things that have already been identified as

1 worth of undertaking.

2 So, in your packets, you have a number of
3 summaries of documents, you have one of the Technology and
4 Privacy Advisory Committee, you have one of Peter Swire's
5 excellent article, which I assume Peter may be touching on
6 later. I wanted to direct you to the one-page, two-sided
7 one which seeks to collect together a number of the prior
8 efforts, and I'm going to use this as an opportunity to
9 focus on some, but not all, in these brief oral
10 comments.

11 The first -- it would appear, now, that there's
12 broad consensus on, is the need for some sort of standards
13 framework to be enacted in some form of law. Now, that
14 could be a statute, that could be a regulation, that could
15 be an Executive Order -- lots of disagreement about the
16 specific form that it would need to take, but that there
17 needs to be some broad level of authorization by somebody.

18 And that the purpose of this is, of course, not
19 merely to ensure that privacy is protected, but to ensure
20 that data mining resources are well-invested, and to ensure
21 that the individuals who were involved in these programs
22 are protected. So that they are given clear guidance, and

1 they know when they stay within that guidance, they are
2 going to be protected by the resources of the government.

3 And the second on this list of 9 points, of
4 course, is that, then compliance with that law or
5 regulation or set of standard or Executive Order matters.
6 That there is really never a justification for saying,
7 "We're going to act outside of the law." And in fact
8 that's true, certainly, of the law we have today, which
9 already -- includes provisions for emergencies, for
10 obtaining authorization after the fact. That to say we're
11 simply going to cast aside the law, is not an option.

12 Some of these, let me go through a little more
13 quickly -- it's important to evaluate effectiveness, there
14 should be ways of doing that systematically, there should
15 be ways of doing that continually, so, repeatedly, not just
16 while the system's being developed, but after it's been
17 deployed.

18 There should be limits on who can use the system,
19 and for what purposes. And therefore, particularly to deal
20 with this subject of mission creep, which we've heard of in
21 many, many settings, but especially over the past two days.
22 There should be clear limits on who and for what purposes

1 the data and the data mining are used.

2 Some form of external authorization for specific
3 data mining programs -- and this is the area, probably, of
4 greatest controversy of the many documents on which this
5 draws. Some would say you have to go to a court, you have
6 to go to either the FISA court, or to an Article III judge.
7 Some would say Congress has to authorize it, or perhaps a
8 special committee of Congress. The Technology and Privacy
9 Advisory Committee used this phrase, "An official confirmed
10 by the Senate," in order, I think, to highlight the point
11 that somebody who was both of a high enough level, and of
12 enough, sort of, political sensitivity, so that the -- we
13 did not have the situation in which the same person or
14 group that proposed the system, also authorized it. That
15 we're trying to separate out those two functions.

16 Working down this list, six, the use of some of the
17 technologies that were talked about in the last panel, that
18 we've talked about yesterday, as well. Data minimization,
19 anonymization -- other technological tools to try to make
20 the data mining compliance with the standards or the
21 framework as efficient and as consistent as possible.

22 The use of audit tools, both to allow for greater

1 compliance during the data mining, but I think, in
2 particular, to create a record for going back to resolve
3 incidents, after they have arisen.

4 Eighth, a system of redress. And the redress
5 point is worth spending just a moment on -- redress here
6 really implies -- includes two separate concepts. One is
7 that the individuals who were affected have a chance to
8 have that impact assessed, and to be -- if not compensated,
9 which would suggest financial compensation -- at least to
10 have whatever benefit or service they've been denied, to
11 have that now provided. So, if you're denied boarding the
12 aircraft, so that the next time you can board the aircraft.

13 Redress to ensure that we don't continue to
14 affect the same individual in the same way, but also
15 redress from the other side, so that the system learns from
16 its mistakes. So that process of constant iteration that
17 we talked about yesterday is built into the system. And
18 this is one place where, I think, frankly the public is
19 astonished and confused in the ways in which we interact
20 with most systems today, having to deal with security,
21 that, you know, we pull a bag because something set it off,
22 but do we learn anything from that experience? Do we

1 actually improve our systems, or do we just say, "We're
2 going to continue to make this mistake," every day, and
3 that redress is intended to help deal with that, as well.

4 And then finally, and in some ways, perhaps most
5 important, the notion of some form of serious independent
6 and rigorous oversight of data mining. And this, again,
7 would be accomplished at many levels -- at the agency
8 level, it would be accomplished through agency Inspector
9 General investigations, which to date, have probably been
10 one of the most revealing forms of oversight we've had --
11 and also by Congress.

12 And, in fairness, I think it is appropriate to
13 say that all of these various reports and documents that
14 have looked at the type of oversight that exists today,
15 have tended to be very critical of Congress for its failure
16 to provide either systematic or useful oversight. For
17 example, to identify a single committee in each House to
18 exercise the oversight function, or to exercise consistent
19 oversight, so that it's not within the same month
20 compelling -- commanding DHS to engage in data mining, but
21 then prohibiting the Department of Defense from engaging in
22 data mining. It's very difficult to figure out this type

1 of inconsistency. And frankly, for the people engaged in
2 data mining, it creates an untenable situation.

3 So, I think these 9 by no means establish the
4 full range of things that might be undertaken, but they
5 establish 9 broad standards where there is, I think, quite
6 a large consensus that these are important -- I think many
7 people would say these are necessary -- and in this case,
8 it's not a question of needing to invent something new,
9 it's needing to do what we already know needs to be done.

10 And let me stop there.

11 Ms. Toby Levin: Okay, one could say that those
12 were a high level, those are high level principles for
13 consideration. So, now I want to ask the panel, are there
14 other considerations that were not included in those nine?
15 Or can we now bring it down a few thousand feet to
16 articulate some more specifics that might be helpful to
17 guide the work of this Department?

18 Peter?

19 STATEMENT OF PETER SWIRE, C. WILLIAM O'NEILL
20 PROFESSOR OF LAW, MORITZ COLLEGE OF LAW, OHIO STATE
21 UNIVERSITY

22 Mr. Peter Swire: So, this is the last panel, and

1 I've really enjoyed being here all day yesterday, and so
2 far today, and feel like there's been a bunch of learning
3 that might help inform what I'm going to say.

4 Yesterday, David Jensen and I, in the question
5 session, had disagreements about whether subject-based
6 versus pattern-based was a useful distinction. And we got
7 to continue the discussion at dinner last night, and I
8 think we came to an agreement.

9 So, at least, I'm going to try to describe it, or
10 at least a way to -- he's at least nodding somewhat, and
11 willing to let me go on for a few sentences.

12 [Laughter.]

13 Mr. Peter Swire: The very excellent point that
14 David made yesterday was that at some level, subject-based
15 and pattern-based are both just different data points,
16 different sorts of evidence that get used for your overall
17 conclusion, and so really, this distinction that's been
18 used often about data mining isn't so helpful.

19 But, I think at dinner we came to, at least -- or
20 at least I'll say it my way and then in the question
21 session, he can say it later -- so there are some things
22 that are very much subject-based, which is the way we've

1 traditionally done law enforcement.

2 I get a hot tip that Danny Weitzner downloaded a
3 song on the internet, that's an important social thing to
4 stop at all costs, and so we're going to then go after that
5 individual subject, develop probable cause to grab his
6 laptop, and do the important things that --

7 [Laughter.]

8 Mr. Peter Swire: I know, you don't want to lose
9 your laptop, that's a different meeting, we're not doing
10 that one today -- anyways, but that's a subject-based
11 moment. I had a very identified thing about Danny, and
12 that led to probable cause, reasonable suspicion, something
13 like that.

14 On the other hand, other end of the spectrum, you
15 have pattern-based. I have no names at all, I just have
16 for downloading songs, somebody who has a Facebook
17 membership, somebody who has six other attributes, it turns
18 out, when you do the data mining, that certain people at
19 the end have a high likelihood of having downloaded songs.
20 No names. Those are ends of the spectrum, those are quite
21 different.

22 You end up, though -- there's some parts in the

1 middle that are a little different. So, it could be that -
2 - to pick on Danny just a bit more -- that he's had emails
3 and other web traffic with certain people, and those people
4 have a heightened level of suspicion, so that looks sort of
5 subject-based, because it has to do with Danny's network.
6 But it fits some overall pattern of when we get suspicion,
7 so that looks pattern-based. And so we have not the end of
8 the spectrum, but some things in the middle that have
9 elements of both.

10 And this is similar to another famous distinction
11 in privacy, between personally identifiable information --
12 Danny's name, address, phone number -- and things that are
13 de-identified -- no idea who Danny is -- and right now,
14 there's a lot of things in between. Like, your IP address
15 is in between and there's big policy debates -- is that
16 identified or not?

17 So, again, you have from identified to de-
18 identified, you have from subject-based to pattern-based,
19 and a bunch of cases in the middle of the spectrum where
20 it's a little fuzzy.

21 But we can usefully say, at the end of that, that
22 some things are subject-based -- and that's what law

1 enforcement traditionally focused on a lot -- and then some
2 things are pattern-based, and I think when you look at the
3 discussions about data mining, having lots and lots of
4 pattern-based -- unless we have a bunch of framework in
5 place, has led to problems, and suspicions and concerns.
6 And so this distinction that -- yesterday there was an
7 argument, "Is this even a useful distinction?" Instead, I
8 think, can be understood as this spectrum between very
9 subject-based, very pattern-based, and we do know that
10 there has been at least some extra level of concerns about
11 the dragnet quality and other problems that come up with
12 pure pattern-based.

13 So, I'll stop there.

14 Mr. Daniel Weitzner: So, since -- oh, go ahead,
15 Barry.

16 STATEMENT OF BARRY STEINHARDT, DIRECTOR, ACLU
17 PROGRAM ON TECHNOLOGY AND LIBERTY

18 Mr. Barry Steinhardt: I hate to interrupt you,
19 Danny, especially since your name has been taken in vain,
20 here.

21 [Laughter.]

22 Mr. Peter Swire: And his computer.

1 Mr. Barry Steinhardt: And your computer is just
2 hanging by a thread.

3 Mr. Daniel Weitzner: You'll protect me if my
4 computer is actually --

5 Mr. Barry Steinhardt: You know where to come if
6 your computer is actually seized, you see me.

7 Let me say that , also, that I have greatly
8 enjoyed this meeting, I thought it was very useful. I just
9 wanted to add a couple of things.

10 I thought Fred's summary was very, very helpful,
11 I want to add a couple of things to the summary that I
12 think we need to talk about.

13 One is the Privacy Act, which I think is in
14 desperate need of modernization. And I know that there are
15 some members of Congress who -- some of them fairly high
16 ranking -- who feel the same way, and I hope the next
17 Congress will take a look at that.

18 There are two things, in particular, that I would
19 note here. One is -- and a special resonance on this
20 subject of data mining is the third-party doctrine, and
21 this notion that if the government collects data from a
22 third party, it doesn't collect it itself, but it simply --

1 goes out, for example, to the free market and buys it --
2 that somehow the Privacy Act does not apply. Even the
3 procedural, or the notice requirements of the Privacy Act
4 do not apply. That needs to get dealt with.

5 And secondly, we have to come to grips with
6 what's a routine use under the Privacy Act, both for the
7 transfer of data outside of agencies, but I think also, the
8 Privacy Act needs to reflect the fact that we now have
9 agencies like the Department of Homeland Security which has
10 lots and lots of functions, and lots and lots of
11 subdivisions, and the Privacy Act needs to reflect that.

12 Thirdly, I do think that we need to examine the
13 question of who is going to be the watchdog here, perhaps
14 who enforces the Privacy Act. We need to come to an
15 international consensus on this issue, and have some form
16 of independent official -- a Privacy Commissioner, or a
17 Privacy Commission, or perhaps a much-enlarged Privacy and
18 Civil Liberties Oversight Board, if it ever gets populated
19 and exists, that looks at those issues. So, I do think
20 it's useful to talk about that. I'm not sure whether
21 that's at, you know, 5,000 feet or 20,000 feet, but I would
22 just add those to the mix.

1 The other issue which I -- which perhaps is a
2 little closer to the ground, here, that I would just
3 highlight is this question of research. You know, one of
4 the things I think that I came away from the last couple of
5 days is that we need to come to grips with the issue of
6 research in data mining and how that can go forward in ways
7 that don't infringe on -- get us the results that we need
8 to get in order to make judgments about whether this is
9 data mining, that any particular data mining proposal is,
10 in fact, useful, has some use in crime fighting, or anti-
11 terrorism, before we move forward with it.

12 But, at the same time, it's cabined enough that
13 it's understood that it's, in fact, research and not
14 operation, and we certainly -- I can't speak for the ACLU
15 at large, or the civil liberties community at large, but
16 certainly would be interested in talking to DHS and others,
17 about how to make sure that we can cabin the research in a
18 way that's understood, it's that there's an opportunity to
19 both get the fruits of the research and to have a public
20 policy discussion before it becomes operational, but to
21 allow the research to go forward.

22 As I said yesterday, I think that the first

1 question that we always need to ask when one of these
2 proposals is made is, will it, in fact, make us safer?
3 Will we get any significant security gain from the
4 proposal, because if we don't get any significant security
5 gain, any enhancement of our safety out of the proposal,
6 then there's no point in even debating the question about
7 whether or not it intrudes on our civil liberties -- why go
8 forward at all?

9 There are times when research is necessary to try
10 to begin to come to grips with that question, and we need
11 to figure out a way to do it.

12 One model that was mentioned yesterday that's
13 probably worth exploring here, was the Institutional Review
14 Boards, the IRBs that exist in the medical context. That
15 may be a -- it's a fairly formal, rigorous process, and it
16 may be a model that we want to look at in this field, as
17 well.

18 Mr. Daniel Weitzner: So, I also like Fred's 9
19 points, I want to just amplify on one, which is this
20 question of system improvement, feedback, learning.

21 I think that one of the baseline privacy problems
22 that we have here is that we don't have a good handle,

1 either from a, sort of expert public policy perspective,
2 from a technical perspective, or from a perspective of
3 public perception, on just how powerful these data mining
4 tools are. We just don't have good ways, technically, to
5 characterize the intrusive or revealing power.

6 And I think most importantly, we -- if you kind
7 of line up our current experience with data mining with the
8 historical experience of wiretap and electronic
9 surveillance, which is another arena in which we had to
10 have a long-running debate about how much intrusive power
11 we want the government to have, and for what purposes -- we
12 were dealing there in the area of electronic surveillance
13 with really a fantastically simply technology. It was
14 just, you know, "Give me a copy of what you already have."

15 I think that -- that this really goes to Peter's
16 question about this spectrum between subject, or -- versus
17 pattern-based data mining. I think the reason we get
18 concerned when we leave that hard edge of the subject-based
19 research is that we start to talk about the application of
20 intrusive power that we don't know how to quantify, we
21 don't know how to understand what it is.

22 And so to the -- people raised questions earlier

1 about the fact that, you know, we want to have the most
2 analytic power available, of course, for public health
3 research, because we want to find the disease carrier, or
4 we want to find the terrorist who's getting onto the
5 airplane.

6 But what we don't really know -- we don't have a,
7 I think, an intuitive, or a public policy sense of just how
8 much power we're putting in the hands of the investigators,
9 so as soon as we depart from the -- the sort of traditional
10 model of an investigation that's going after a single
11 person, or a set of -- a limited number of years -- as soon
12 as we amplify the power of investigation with this data
13 mining technology we get, I think, appropriately worried.

14 We worry, partly, about effectiveness, but I
15 think more importantly, we worry about power. We worry
16 about what power we're putting in the hands of government.

17 And I think that -- I think we're going to have
18 to feel our way along this spectrum, because I don't think
19 this technology is going away, it's been used in the
20 private sector, as I think was probably discussed, for 20,
21 30 years. I thought one of the neatest tricks of the TIA
22 program was to pretend it was something new, as opposed to

1 just something that was going to be applied in a new
2 context -- and I would say that because of that, I think
3 it, it really puts a lot of -- it requires that we put a
4 lot of stress on this, kind of, system improvement and
5 feedback tab, and we have to work as hard as we can to make
6 the uses of data mining as transparent as possible. There
7 are, obviously, security limitations on being able to do
8 that.

9 But, again, to come back to the wiretap example,
10 we reached some sort of reasonable, you can argue, but we
11 reached some social middle ground, some policy middle
12 ground, about what kind of electronic surveillance was
13 going to be acceptable, and what not, by going through, you
14 know, hundreds and hundreds of cases in open court, and
15 sort of teasing out the question of the balance of power
16 between individuals and government.

17 And I think we have to figure out how to have a
18 similarly open process for the application of this new kind
19 of intrusive technology, otherwise, we're never going to be
20 able to answer the sort of basic questions about who ought
21 to have what kind of rights, and who ought to have what
22 kind of power.

1 STATEMENT OF THOMAS OSCHERWITZ, VICE PRESIDENT
2 OF GOVERNMENT AFFAIRS, AND CHIEF PRIVACY OFFICER, ID
3 ANALYTICS, INC.

4 Mr. Thomas Oscherwitz: You know, I agree with
5 Fred's 9 principles so much that I'm actually sharing his
6 mic --

7 [Laughter.]

8 Mr. Thomas Oscherwitz: But, I think one way to
9 actually approach this problem, I think is to actually
10 unpack the word data mining. If you ask yourself the
11 question, what should we do about employing protections
12 against data mining or for data mining, you get certain
13 reactions, I think, often because it's conflated with
14 surveillance.

15 But taking a definition that Professor Jensen
16 mentioned yesterday, what's your response if the question
17 is, "What are your feelings about an iterative process of
18 learning and probabilistic inference about interconnected
19 data records?" It's just a different connotation when you
20 think about it in that perspective.

21 And the reason why I say that is, when we're
22 talking about data mining, we're actually packing several

1 things together. And one of the points that was made
2 earlier in the conference, or workshop, was that there's a
3 collection process, there's an analytical process, there's
4 what happens after you do that analysis?

5 And so, from the perspective of talking about
6 data mining we need to recognize that it's a technology,
7 and some of the concerns we have are independent of the
8 analytical process.

9 I'd also like to point out -- going back to the
10 Professor's definition, some of the words, although I can't
11 pronounce them -- iterative. The implication there is that
12 when you're evaluating issues like efficacy, it's
13 iterative, so you have to be able to look at efficacy over
14 time. Efficacy is going to change. So, when you make an
15 analysis of a program, we have to recognize how the
16 technology works.

17 The second point is probabilistic inference.
18 That deals with issues about what you do when the model is
19 implemented. If you -- this is not a yes or no answer, so
20 when you're evaluating what to do with this information,
21 how are people who are going to get this information going
22 to use that information, and what are the consequences

1 based on the probabilistic inference?

2 The other -- I would point out is, not all data
3 mining is the same. If we're talking about an analytical
4 process, even the definition used by Congress gives an
5 exception for fraud. You're using data mining, but you're
6 using it for fraud. And the question about whether or not
7 external authorization and other types of controls -- I
8 think it really goes back to, what are you using the
9 information for? Is it a new use of information? What are
10 the consequences of that information?

11 So, I'm not sure if it's a one-size-fits-all
12 legislative or regulatory solution for data mining, it also
13 depends on what the information is being used for.

14 Ms. Toby Levin: Well, I think building on the
15 elements that were on Fred's list, and some of the things
16 that have been added, what we're really looking to do is
17 develop a set of considerations, a set of principles to
18 consider when we're conducting data mining research.

19 And let me just add some complexity to that
20 challenge is -- several I guess, Danny mentioned
21 specifically, the need for making the data mining project,
22 the activity, more transparent, as transparent as

1 possible -- well, how do we do that when some of the
2 activities may be in a classified setting? How do we take
3 these principles that we're working on drafting, and deal
4 with the fact that some projects may be classified?

5 Mr. Daniel Weitzner: I think that -- I want to
6 just connect that question to Tom's point. I -- as you
7 would have heard from the last panel -- I very much agree
8 with the view that we do have to focus more attention on
9 the ultimate uses of this analysis. That's where, I think,
10 some of the important new legislative and rulemaking
11 activity is going to have to come.

12 As to the question of how we look into what will
13 often be a classified, or at least a sensitive,
14 environment, I think that there are two ways to approach
15 that question.

16 Number one, Fred mentioned the role of Inspectors
17 General, particularly, the DOJ Inspector General has been
18 able to uncover quite a bit of information about what's
19 been going on, I think that process ought to be able to
20 continue.

21 So that even if any -- if a given data mining
22 activity, whether it's for research or for operational

1 purposes, even if that's happening in a classified
2 environment, there ought to be enough information preserved
3 in that classified environment that enables some third
4 party to go in and evaluate both the effectiveness and the
5 intrusive power of those systems.

6 And that information -- the results of that
7 information, ought to be made visible, to regulators and
8 legislators, and ultimately to the public.

9 Ms. Toby Levin: But that's after the fact,
10 Danny. What about before? What about the early stages?

11 Mr. Daniel Weitzner: Well, I think that can be
12 ongoing. I think that tests of systems can be subject to
13 that same kind of analysis. If systems are going to be
14 tested with synthetic data, I think it's actually a great
15 place to start testing what kind of oversight process could
16 work. And it may be oversight over the privacy interests
17 of dummy data subjects, but that's not a bad way to start.

18 I think that there is a lot that can be analyzed
19 in a way that doesn't reveal sensitive information that
20 still tells us things that we need to know about
21 effectiveness questions, and about ultimate intrusive
22 power.

1 But again, if we don't start by expecting a high
2 level of transparency of these systems, even if the
3 transparency is only to people who have the appropriate
4 clearances, then we're never going to get past this kind of
5 shadow boxing debate, where we don't even know what we're
6 talking about, we don't even know what kind of intrusion
7 we're actually trying to regulate, or told we shouldn't
8 worry about.

9 So, I think that's where we have to start.

10 Ms. Toby Levin: Okay, Barry?

11 Mr. Barry Steinhardt: I think the real problem
12 here is the overuse of various secret classifications,
13 which has become, over the last 7 years, sort of endemic.
14 Where, virtually everything is now characterized as
15 "sensitive security information."

16 I've had some experiences that border, kind of on
17 the serial comic. There was a meeting, I don't know, about
18 a year ago to talk about the issue of the fairly aggressive
19 pat-down searches that the TSA was then doing, particularly
20 on women. And, you know, in a moment when I thought I was
21 trying to be helpful, you know, I asked the question, well,
22 you know, "When are you going to move towards some form of

1 technology that allows you to detect things other than
2 metal? After all, you know, we all go through the metal
3 detector now," and I was stopped and I was told, "Well, no,
4 we can't tell you whether or not the metal detector only
5 detects metal. That is sensitive security information."
6 This, by a high-ranking member of DHS.

7 And we went back and forth on this for about 10
8 minutes, how, sort of Alice in Wonderland all of this was.
9 But, the point was, every time we, you know, we try to deal
10 with these issues, we run up against a claim that something
11 is sensitive security information, and therefore cannot be
12 discussed -- whether it's how many people are on the watch
13 list, or whether the metal detector can detect things other
14 than metal. And, you know, we can't even have this
15 discussion until we have a more transparent government.
16 And until we have a more transparent government, it's not
17 possible to say that we're going to have a real public
18 policy debate about whether technology is worth using or
19 not, or data mining program is worth using.

20 Mr. Peter Swire: So, Toby's question was, in
21 response to the demand for transparency, what about the
22 fact that so many of these things are classified or

1 sensitive? I had three quick points about that.

2 The first is, if it's going to be a system that's
3 in the classified space, you still want to have, what I've
4 called, due diligence, by somebody who's not the proponent.
5 So, you have the enthusiast for every project in
6 government, or in a company, and then you want to have
7 someone else who doesn't have a stake in having the program
8 win, at least look at it.

9 And that may mean something that I think was
10 briefly mentioned in some of the materials here -- there
11 might be a supplementary Privacy Impact Assessment when it
12 comes to data mining. You know, you might have your
13 standard PIA, and then it might be there's an additional
14 set of questions that get built up over time.

15 The second point is, what would those additional
16 set of questions be, and on this, I'm a little bit of an
17 optimist. So, I think that there's been a social
18 learning curve about data mining. The list of footnotes in
19 Fred's document shows just a huge outpouring things,
20 compared to what we had three or four years ago, about
21 what the criteria are for data mining -- what are the
22 issues, what are the legal structures. And then, the fact

1 that this room is as full as it is today on privacy and
2 data mining, when people sat here for 2 days -- there's a
3 lot more people who know a lot more than they used to about
4 these issues, and the Privacy Office at DHS and all of the
5 other organizations you represent can come into play, here.
6 So, I think we have a much bigger set of things we know
7 than we did.

8 The third point is that audits -- as Brian
9 Tretick, I think, said earlier today -- are not just for
10 after-the-fact detection. The knowledge that everything's
11 going to be audited is supposed to prevent misuse -- it
12 sends a signal to people that you're supposed to follow the
13 rules, and it's supposed to, going forward, have closer
14 approximation to what we're supposed to do.

15 Mr. Fred H. Cate: Toby, I just wanted to add the
16 somewhat mundane point to your question, and that is, in
17 some ways, the classification issue has become entirely a
18 way of, I think, misdirecting the debate. Because we have
19 so many tools for dealing with oversight in classified
20 environments, and it's not like we're using all of them,
21 and then saying, "Now what?" We're not using them.

22 So, in other words, do we have a Privacy and Civil

1 Liberties Oversight Board -- which is supposed to have a
2 security classification -- you know, Congress won't even
3 consider the nominations for that. I mean, we don't even
4 have it. It's not -- it doesn't take a genius to say that
5 would be a useful place to start for oversight -- we have
6 internal Inspector Generals who, in many instances, have,
7 you know, been extremely important in bringing about
8 effective oversight -- they operate in a classified
9 environment, we have the Privacy Office within DHS, which I
10 assume, also operates in a classified environment.

11 We have members of Congress and committees of
12 Congress with clearances who operate in a classified
13 environment. I think on this, the public is really quite
14 understanding. You know, I understand that you're not
15 going to be able to tell me, or publish in the newspaper, a
16 lot of the details about the activities in which the
17 Department's engaged.

18 It is the apparent unwillingness to tell anyone
19 about it that is the agonizing issue. And I think, exploiting
20 those very, already widely available tools would then make
21 it much easier to say, "But we're not telling you."

22 You know, one of the recommendations of the TAPAC

1 was to have external advisors who would have clearances, so
2 that you would take the year-long process of getting them
3 their clearance, they would be ready, and then when you're
4 developing the program, and you need to bring Danny in, or
5 you need to bring somebody in to get a read on this, they
6 would already have gone through that process and be
7 available.

8 You know, I know that that's done in limited ways
9 today, it ought to be done, in some ways, in more visible -
10 - you know, the ways that corporations advertise, we have
11 Consumer Advisory Panels, you would think agencies would be
12 advertising, "We have the very best privacy advice that we
13 can get," even if we had to get it in a classified
14 environment.

15 Mr. Thomas Oscherwitz: Just to, sort of echo
16 Fred's point, I do think there's --

17 Mr. Fred H. Cate: This is my microphone --

18 Mr. Thomas Oscherwitz: Thank you, Fred, that's -
19 - yeah.

20 Just to echo what Fred said, I do think some of
21 the benefits that in a public forum you have in terms of
22 expert advisors, or IRBs, that was mentioned by Barry

1 before, or advisory committees -- some sort of expert
2 process would also be beneficial here, especially in terms
3 of issues like efficacy. You know, is this program
4 actually going to provide any valuable, or useful, or
5 possibly useful results?

6 Having external advisors not only on privacy, but
7 on whether data mining, in this context, is actually
8 effective, I think would be quite useful, regardless of
9 whether it's a public or private context.

10 Ms. Toby Levin: Okay, so I think all of that
11 went very effectively to addressing the question I posed.

12 Let me ask now what we can learn from private
13 sector experiences here with regards to other regulatory
14 models that may be useful in looking for protections, or
15 processes or redresses.

16 Tom?

17 Mr. Thomas Oscherwitz: Sure. I think I'd like
18 to make a couple of points here. Going back to that
19 definition that I couldn't pronounce, about iterative
20 processes, in terms of making probabilistic inferences --
21 one of the things, you know, the private sector has had a
22 lot of experience in is that when you're building a data

1 mining model, the first model may not necessarily be the
2 most effective. There is a process where R&D is crucial.

3 And in terms of going forward, there does need to
4 be some space, and it's recognized in some statutes, for
5 research and development. So, one point I would make is
6 that there needs to be some sort of separate process for
7 allowing folks to build the models.

8 Along those lines, data acquisition and model
9 building is different than data use and model
10 implementation. When you're looking at building a model,
11 you need to have access to a broad array of resources,
12 because you may not know, necessarily, what works and what
13 doesn't.

14 So, one insight that, I think, we should consider
15 is restrictions on access to information, building the
16 model and the research process may actually limit the
17 effectiveness of the model.

18 On the other hand, I would also point out that if
19 you look at various statutes like the Equal Credit
20 Opportunity Act, there are societal considerations, even in
21 the model-building process itself, which is, there's a
22 determination made that certain variables are simply off

1 limits, because the incremental advantage of using those
2 variables is outweighed by the detrimental impact and other
3 aspects of society, for example, invidious discrimination
4 based on race or religion.

5 And so, there may be a couple of variables that
6 prior to actually building the model, even though I'm
7 generally advocating for broad access to information, where
8 there may be some limits, and there could be some advice
9 there.

10 So, one thing we have to think about, you know,
11 from the private sector experience is that there may be some
12 variables that need to be off limits, but generally keep it
13 broader.

14 The other thing I would say is that rules
15 actually do work in the private sector. There are fraud
16 exceptions in a variety of statutes, which allow data
17 mining in that context, and you said data mining in that
18 context has been quite effective, but it's also limited in
19 how you can use that data. So, the private sector, I
20 think, has a lot of experience with purpose limitation, and
21 that's been, I think, a very effective way to deal with
22 some of -- what I consider the biggest risk of data mining,

1 which is mission creep. Which is, you start with one
2 purpose, and then the purpose expands, and expands and
3 expands.

4 There's been a number of discussions about
5 building design specs in the model that respect privacy --
6 I think that's very critical. One of the things that we
7 often talk about in the private sector is something called
8 investment risk, where you build a model, and then the
9 rules change in the middle of the game, and all of a sudden
10 the model that you built is no longer applicable, because
11 certain data elements are no longer available.

12 So, to some degree, we may not necessarily know
13 what the privacy concerns are going to be 5 years from now,
14 but you want to have a system that can adjust and say, wait
15 a second, this variable is no longer appropriate. I'm
16 thinking here, for example, of the Driver's Privacy
17 Protection Act, and how folks were using that data at one
18 point in the late nineties, and no longer being used. So,
19 that would be the other point I would make on access to
20 information.

21 I guess the last point I would make is, going
22 back again to the technology itself, which is that, you

1 know, data mining is policy neutral. And one of the things
2 that you have to do when you're evaluating data mining is
3 compare it to the current system that's being used. You
4 know, airline scanners are already screening passengers --
5 don't we want them to improve their criteria for screening?
6 Cargo is currently being screened, don't we want to improve
7 that process?

8 And so, one of the points I would make about the
9 private sector is, we've used data mining in a lot of
10 different contexts to improve current processes, and that's
11 a lot of what this data mining is about.

12 Mr. Daniel Weitzner: Could I make one quick
13 response? I think it's a mistake to say that data mining
14 is policy neutral. I think that it is -- and by that, I
15 don't mean that it's bad -- but I think data mining brings
16 a qualitatively different set of regulatory requirements to
17 the table. We can't expect to have accountability to
18 whatever set of privacy and civil liberties rules in data
19 mining by using the same techniques that we have for the
20 cop investigating, you know, a murder on the streets, or a
21 drug dealer. And it's because of the scale problem.

22 And I think you'd probably agree with this, and I

1 may be taking your definition of neutrality differently
2 than you mean it, but I think that we can't -- we have to
3 have kind of a parity between the intrusive power of the
4 tools, and the accountability power of tools on the other
5 side.

6 That we -- in that the whole idea of data mining
7 is, in some sense, to enable investigators to discover
8 things they wouldn't otherwise discover by application of
9 their brain, we need enforcement tools that enable
10 enforcers and regulators to discover violations that they
11 couldn't discover just with their brain. That's sort of --
12 the problem here is the, sort of, scaling up, and I think
13 we have to make sure that the scaling happens both with the
14 analytic power, and with the enforcement power. It's got
15 to go to together, somehow.

16 Mr. Thomas Oscherwitz: My only response, I mean,
17 I think I -- there isn't that much disagreement at all. I
18 guess I would try to unpack the analytical process from the
19 data collection process where, if you have vast more
20 quantities of information, that definitely changes the
21 calculus.

22 I certainly agree that the ability to provide

1 greater insight does have some implications, but again, is
2 data mining that different from a lot of other tools out
3 there that increase -- I mean, the internet, searching --
4 are we having a workshop on DHS's use of the internet and
5 using search terms? I mean --

6 Ms. Toby Levin: Not yet.

7 Mr. Daniel Weitzner: And I would put those, in
8 many ways, in the same category, so I agree --
9 -- I don't think that -- I don't think what I'm
10 saying applies narrowly to data mining, but I think it
11 applies to this question of the scale of investigators'
12 powers has got to be matched by the scale of regulatory
13 response.

14 Ms. Toby Levin: Okay.

15 Mr. Thomas Oscherwitz: I want to go back to your
16 question for a moment, Toby, sort of what can we learn from
17 the private sector -- one of the things I think that we've
18 learned from the private sector over the past few years, is
19 the degree to which the private sector has been, in a
20 sense, compromised by the government. By which I mean that
21 the private sector collects all sorts of data for all sorts
22 of purposes, sometimes under fairly heavy regulation,

1 restrictions of what can be done with that data, what can
2 be collected, et cetera.

3 But, all of the laws -- virtually all of the laws
4 -- that govern the collection of that data have these huge
5 exceptions for law enforcement or national security
6 exceptions. And I have to be looking at Peter, here, the
7 very first conversation Peter Swire and I ever had was
8 about the law enforcement exceptions in the HIPAA
9 regulations.

10 Mr. Peter Swire: A highly specialized
11 conversation.

12 Mr. Thomas Oscherwitz: Highly specialized, but
13 foreshadowed many, many future conversations with Peter.

14 But, in general, we need to go back, and I think
15 we need to go back to kind of first principles here, which
16 is to say that, you know, even to the extent to which we're
17 going to regulate the private sector and say to the private
18 sector, "These are the rules that you must follow to
19 protect privacy," we cannot have these huge holes for law
20 enforcement, or national security acts, as we have to go
21 back to the principle that there's got to be a predicate to
22 get that information, in most circumstances, you've got to

1 go to a neutral arbiter or magistrate to get that
2 information -- we've got to go back to that, or all of the
3 rules we can create for the private sector get
4 swallowed up by this exception.

5 Mr. Peter Swire: I'm not going to talk about
6 HIPAA and national security, but I have some comments on
7 research, because I think what Tom said and -- it's been
8 interesting -- so Barry, earlier in the panel today said
9 we've had some interesting promise here that maybe we can
10 have a research approach with Institutional Review Boards,
11 IRBs and I think that it may be that a follow-on process or
12 workshop a year from now might be on research and privacy.

13 But, we've seen interest from S&T the last two
14 days, and I have a few comments on it. One is something
15 that Tom said that makes total sense analytically, but I
16 don't think most people have focused on, which is at the
17 research level, the researcher has reason to have more
18 data, rather than less. It's not a little tiny test bed,
19 instead, if you're going to do data mining research, the
20 logic of the research is you want lots and lots and lots of
21 data so you can throw out 90 percent of it and find the 10
22 percent that's most useful.

1 Well, I'll just note that if there's a government
2 program that's collecting lots and lots and lots of data
3 for the first time, that raises a whole bunch of privacy
4 issues, and governance issues, and how's it going to work
5 out. When Jet Blue handed over its research data, it led
6 to a certain level of controversy, so the fact that it's
7 called research doesn't immunize it from controversy.

8 But I think that if we want to have research
9 going forward, and we want to have better tech going
10 forward which, you know, probably all of us do, it suggests
11 that we're probably going to need to develop some
12 institutional things around that, that won't be exactly
13 like the IRBs for HHS. And so, I just offer a few
14 thoughts.

15 One is that the research side will be under
16 enormous pressure from the operational side. And just
17 having been around conferences, around DHS over the last
18 several years, you can imagine a conversation that goes
19 something like this: the researchers are doing the
20 research, they have all of this data and somebody says,
21 "Don't you realize the next attack could come any day?
22 It's urgent, we need to get answers now, we can't wait

1 until this thing's 100 percent done, we have to act now.

2 So, give us what you have now, because we need it, because

3 it might be the only way to stop the next attack." Right?

4 I think a lot of us recognize some variation of that as an

5 important part of the urgency to have Homeland Security to

6 stop the next attack that we've been living under.

7 If that's the case, then -- when we create, as

8 we create the research side, we're going to need to think

9 about what the research folks say to that urgency request.

10 One possibility is that there's a huge firewall,

11 and the answer is never, ever under any circumstance --

12 that will help us do more research, if we say that. But

13 there will be enormous pressure at certain moments to start

14 to get preliminary reports out, or something like that.

15 So, there's going to have to be a whole conversation about

16 when does the research, you know, get accelerated, and move

17 over to the urgency side -- that happens in medical

18 research, right? If you're doing a double-blind study, and

19 it turns out one half of the study people are dying,

20 and the other half people are living, you cut off the study

21 -- you don't watch to see how many more are going to die,

22 right? We're getting a nod from our medical researcher.

1 So, that's something that I don't think has been
2 highlighted up till now, which is when does research cross
3 over into operations, and there's going to have to be a
4 bunch of learning and discussion about that. And the
5 reason is that, the people who don't want the government to
6 have lots and lots and lots and lots of new data, who are
7 scared about mission creep and all the rest, are going to
8 be very worried that that leakage is going to happen.

9 And so, I just sort of flag that as a big issue
10 in the research area.

11 Another thing to point out is that in research
12 and privacy -- something I've been trying to think about
13 over the last period of time -- there's several other legal
14 regimes besides IRBs that are relevant, that I'll just sort
15 of throw out for people to start thinking about.

16 One of them, which was discussed yesterday,
17 briefly, is at the FDA, the Food and Drug Administration,
18 when you go from Phase I to Phase II to Phase III, there
19 are these gatekeepers, there's people who say, "Yes, we
20 have enough data that it's able to go to the next stage."
21 So, there are some decisions made about what's promising
22 enough for effectiveness and safety, so that you get to do

1 the next step, so that's FDA.

2 There's also been a recent report from Berkeley,
3 from the Samuelson Center, about doing research on network
4 behavior -- when people surf the internet and do things,
5 researchers want to say, "Hey, what's going on with
6 people's surfing?" That's sometimes called research, and
7 sometimes that's call wiretapping, right?

8 [Laughter.]

9 Mr. Peter Swire: So -- or advertising, as Danny
10 said -- so that's something to think
11 through, so you can look at the Samuelson Center's
12 research.

13 Another area for research in American law is
14 under the DMCA, the Digital Millennium Copyright Act,
15 there's a security researcher exception under DMCA, which
16 has been often criticized as not a very good one, but it's
17 another place where research has been built into the
18 American legal structure when you're trying to figure
19 things out.

20 And just one more comment and then I'll stop --
21 so we have these different legal regimes to look at for
22 research, we see that separating research and operations is

1 going to be hard, and I'm looking at my notes, here, trying
2 to find the next wonderful point I meant to say, I
3 apologize here, we've all had this happen -- I'm sorry, I
4 can't remember it. So, I'll stop.

5 Mr. Fred H. Cate: Just two comments, one on the
6 research point -- I absolutely agree, and I think Peter's
7 contribution here is extremely important -- this is one of
8 the critical issues, about how do you do research in data
9 mining technologies, and how do you do it in a way that
10 doesn't, either, get you into political trouble, or into
11 legal trouble.

12 I think, frankly, all of us who live in the
13 research environment face this all the time, I mean, most -
14 - much of the research we do involves deception, or it
15 involves wiretapping -- it involves something that brings
16 us into a legal conundrum, and frankly, our own research
17 infrastructure is not that good at dealing with those
18 issues. You know, if it's wiretapping, it's wiretapping,
19 you know, there's no large research exception for, you can
20 conduct surveillance if your purpose is research.

21 I do think the research discussion suggests,
22 though, again why the type of framework that we've talked

1 about earlier, and that many of these prior groups that
2 have looked at this and talked about, still works pretty
3 well. Because you still need some sort of regulatory
4 framework, you know, for most of us it's IRBs, required by
5 Congress, that's why we all use them, not because we dream
6 them up, but because we're required by law to use them.

7 We need some sort of limits on, you know who gets
8 access -- these are the types of questions IRBs ask. We
9 need some sort of authorization before we can go into
10 action, whether that's the FDA as a gatekeeper -- so the
11 same types of things you see in a framework for data mining
12 generally are likely to work in the research environment,
13 as well.

14 The other point I wanted to make, really going
15 back to Danny and Tom's interaction over the question of
16 the scope or tools for dealing with the scope of
17 data mining, you know, one interesting parallel which does
18 not come from the private sector, but the public sector is
19 the IRS. So, the IRS is probably the ultimate data mining
20 organization; it gets this massive amount of data, some of
21 which is, if you will, volunteered by the taxpayer, the
22 vast majority of which is collected from third parties.

1 And then it combines it all together, and it sends out
2 notices saying what you owe that you didn't think you owed.

3 It is required by law when it does that to send
4 out a copy of the Taxpayer Bill of Rights, to give you
5 access, it must have a -- not really a Privacy Office, but
6 it has a Taxpayer Advocates Office, it has to give you
7 information on how to contact that office, so that once --
8 so you get the bad news, you get the, "You've been caught
9 in our data mining framework," and then you get a bunch of
10 other things along with that.

11 And, I'm not even saying it works particularly
12 well, but it is a step -- I mean, these are tools in the
13 right direction, so that at least there's some response for
14 saying, if I think it's a computer error, or if I think
15 I've been misidentified, if I think the rules are being
16 applied unfairly, there is some recourse set out in front
17 of me, even before I make the payment.

18 Mr. Peter Swire: Can I just follow-up? I
19 remember the last point, and Toby was gracious enough to
20 allow me 30 seconds.

21 So, the last point about research is, if we
22 create a research exception, then everyone else will try to

1 have their stuff look like its research. And this is a
2 familiar bureaucratic thing -- if it turns out that being
3 called a turnip gets you money, then you'll suddenly be a
4 turnip lover.

5 And so here's the way it played out for research
6 for HIPAA, when we were writing the HIPAA research rules.
7 So, research -- when you think of medical research, you
8 tend to think about, you know, solving leukemia, somehow,
9 you know, saving the -- coming up with the clinical trial
10 that's going to solve the next disease. When some people -
11 - let's call them pharmaceutical companies -- think of
12 research, they might think of that as, "Let's be -- let's
13 call lots and lots of people at home and see which of our
14 sale pitches gets the most new drug purchases." And that's
15 research -- that's marketing research, but it's research.

16 And so, when you want to have this, it's
17 wonderful, we love it side of research for medical things,
18 a lot of people didn't want to have that enabling a whole
19 new round of data mining by pharmaceutical companies to
20 annoy people at home, and so there was this effort, then,
21 you have to draw this line between what's the good
22 research, or really research, and what are people trying to

1 shoehorn their way into the definition for other purposes.

2 So, just because somebody puts a label "research"
3 on it, doesn't mean it's going to be something everyone
4 decides is a good thing, something else you have to watch
5 for as you define your research rules.

6 Mr. Daniel Weitzner: Just quickly, I think
7 there's a pretty important strategic choice that, in many
8 ways, is presented to the national security community, and
9 those in the government who would use these technologies.

10 There's a way to approach this research question
11 in a kind of a defensive mode that says, "Well, we'll sort
12 of do enough to satisfy people who are grumbling at the
13 edges, and we'll come up with some studies that establish
14 effectiveness and, that's it."

15 And then there's another approach that says, it's
16 actually important to develop a sense of public
17 understanding and trust, here. Not only is it important,
18 presumably, for the users of these technologies to have
19 them be effective, I think that's going to be kind of a
20 self-executing, self-actuating sort of process, but it's
21 also important for people to actually feel good about it.
22 I think that's what's, to a large extent, happened in

1 medical research, people have a sense of confidence that
2 double-blind studies don't go on killing people when they
3 can stop.

4 And though there are arguments about that around
5 the edges, there's the basic trust in the -- at least in
6 the kind of medically driven research community,
7 pharmaceutical company-driven research may be slightly
8 different -- but there's a basic sense of trust that's been
9 established, so that we feel good about public health
10 research, and the public health process, in general, and we
11 give that process, as we discussed earlier, phenomenal
12 latitude, from a privacy perspective.

13 And I think the question here really, is how are
14 we going to generate that overall sense of trust? So that
15 the public, regulators, others are going to accord the
16 latitude that I think is obviously going to be necessary,
17 because this is all -- these are new investigative
18 techniques, it's new technology -- at least this new
19 application of technology, and so I think it's great that
20 you've raised the question. And I hope that it somehow
21 gets raised beyond the level of just what's the minimum
22 requirement here to satisfy those who are pounding the

1 table, but more, what will actually persuade the public
2 that this is worthy of trust.

3 Ms. Toby Levin: Well, that's the area I really
4 wanted to move into next, because those of us who work in
5 the area of privacy have said for years that, you know,
6 "Don't view," when we talk to our customers, whether it's
7 in the private sector or on the government side, or the
8 agency within which we work, "Don't view privacy as an
9 obstacle, don't view it as a barrier, don't view it as a
10 pain in the you-know-what, rather view privacy as a way of
11 gaining trust."

12 So, to what extent can we demonstrate to the
13 researchers that going through the processes that we're
14 discussing this morning, which clearly will be some --
15 time-consuming process. It's not merely a checklist, at
16 least that's not how we want to do things at the Department of
17 Homeland Security, do we want to do something meaningful,
18 how do we explain this to the research community within the
19 Department, and to the senior leadership at the Department
20 that this is worth doing?

21 Mr. Barry Steinhardt: I'm not sure if this is
22 one where you appeal to people's best instincts, or their

1 most self-protective instincts.

2 You know, I mean, in the latter vein, one answer
3 to that is total information awareness program, right?
4 Which is to say, if you don't do it, it's going to come
5 back to bite you in the rear end. And, you know, the
6 Congress is going to get involved, the advocacy community
7 is going to get involved, the press is going to get
8 involved, you're never going to be able to explain what you
9 were doing, and why you were doing it. That may be kind of
10 the real politic answer here.

11 On the -- I would hope, I'll sort of, just tell
12 you what I think is sort of the better instinct here, which
13 is to say, "Look, we all want to do something that's going
14 to be effective, we all want to protect the homeland, so
15 it's important that we do the research, a) to demonstrate
16 effectiveness, and b) to find out if it's effective, how we
17 can do it in the least intrusive way to preserve our way of
18 life, our liberties," now it's known as the good angel, but
19 I don't know how often we get to think about our good
20 angels.

21 Mr. Daniel Weitzner: You know, there's a lesson
22 from human -- from computer interface design, it's kind of

1 an aphorism from Ben Schneiderman, who's really the founder
2 of the field that looked at what are -- how do you build
3 good interfaces to computer systems that people can
4 actually use and understand, and -- Ben has a
5 bunch of principles, just like Fred's principles, but --
6 the first principle that Ben has is he says, in any
7 computer system, the interface to any computer system, if
8 people can't see a function, they won't use it, and they
9 won't understand it.

10 And I think that, it's a real challenge here, but
11 we talk about issues like redress and feedback and
12 correction in systems, and I think someone was saying
13 earlier, the problem that we have with a lot of these
14 threat detection systems is that people -- the average
15 person's interaction with these systems is really bordering
16 on the absurd. And I think that the challenge that these
17 systems are going to face, is that people, overall, are
18 used to a higher and higher degree of information access,
19 and information interactivity -- we're used to that from
20 our banks, we're going to get used to that, I think, more
21 and more from the medical system, we're used to having a
22 much -- a dynamic relationship with information systems

1 that we depend on -- you can pull them up on the web, you
2 can -- in the best case, you can correct things, you can
3 fix things -- sometimes they're frustrating, but there's an
4 interactivity.

5 The problem with these systems is they seem to be
6 buried behind many layers of protection, and there are some
7 good reasons for that, but I have a feeling that in the
8 long run in order to build a sense of public trust, that's
9 going to have to change. Because it's going to become more
10 and more -- these systems are going to become -- are going
11 to seem more and more anomalous, and frankly, more and more
12 threatening to the extent that you can't even go to an
13 airport and find out why you keep getting stopped for a
14 secondary screening.

15 I don't want to minimize the challenge of doing
16 this -- the IRS is a great example, I think. The IRS, I
17 think, finds over and over again, that they get increased
18 compliance and more effective processing of claims when
19 they're more helpful, provide more data to taxpayers, and
20 make the process easier. And that ultimately makes people
21 feel better about a system they're part of, as opposed to
22 alienated from it.

1 And I think that in and of itself, is a subject
2 for research to ask, how can you accomplish these goals in
3 a way that still provides people some reasonable amount of
4 feedback?

5 Ms. Toby Levin: Well, yesterday we really
6 started to get into the complex issue of data collection,
7 and the many different sources -- private sector, public
8 sector data, data that was collected for one purpose, but
9 now is being used for another purpose, and the issues,
10 then, of building trust when you have such complex data
11 inputs. And it isn't possible, really, to go back to the
12 individual and say, "You know, the information you gave me,
13 you know, 3 years ago for a particular purpose," maybe it
14 was FEMA, you're a Katrina victim, and we collected
15 information in order to provide you with a benefit, you
16 know, do we need to go back to individuals in order to let
17 them know, "Now, we're repurposing your information and
18 we're using it for, in order to develop data mining models,
19 or to develop -- or to be used in another context." It's
20 quite complicated.

21 Mr. Peter Swire: Well, that's where HIPAA is --
22 offers you some hope, right? So, one way you can do

1 research under HIPAA is by consent of the patient. Another
2 way is you can get the IRB to approve it, and you don't
3 need individual consent. And so -- and the reason is, that
4 there's this process that exists where -- with what you
5 hope are expert people -- and that goes back to the
6 question a moment earlier, which is, how do you explain to
7 researchers why they should go through the pain of filling
8 out these forms, or whatever it is?

9 And I think part of the answer is what Barry
10 said, which is, if you don't you'll get your whole program
11 blown up and defunded by Congress.

12 But, I think there are somewhat subtler, and
13 additional, points to make to that -- that's a good one to
14 get people's attention. Basically, you say, how are you
15 doing at getting the really, really big databases today?
16 And the researchers will say, "Well, we've proposed 5 and
17 none of them have been approved," right? And so the path
18 forward for research databases, is probably to have a
19 process with guarantees around it, and then once you build
20 those processes with guarantees, you hope that the IRB or
21 its equivalent becomes a source of expertise -- it helps to
22 shape what the researchers ask for, so they know if they do

1 it this way instead of that way, they're likely to get
2 approved, so you start to build up some experience of what
3 works. And if they see, over a period of time that this is
4 a path to getting things approved, then that's a much
5 happier outcome than having this risk that the whole thing
6 will suddenly become a firestorm of criticism.

7 Ms. Toby Levin: And I think you probably would
8 agree that once you lose the trust, it's very hard to
9 regain it. So if you've got, you know, a lineup of
10 projects and you lose the trust on the first one out of the
11 gate, it's going to make it hard to get the other ones out.

12 Mr. Thomas Oscherwitz: And just to sort of
13 follow-up on what Danny said, I do think there is a value
14 in both the implementers and the public having a better
15 understanding of the program. And I'm just thinking of an
16 example here, where you have a -- predict a model where
17 people are flying because on one airplane 5 people bought
18 one-way tickets. And what that suggests, if you understand
19 the model is that, you know, you can have a false positive
20 that doesn't mean the model's not working, it just
21 identified an outlier. So, the person who's implementing
22 that model would say, "Well, I have no idea whether you

1 have a perfectly legitimate reason for being an outlier, or
2 you're a bad guy, but I'm going to check you out," and the
3 person on the other sides needs to understand, "Well,
4 there's nothing about my innocence or guilt that's causing
5 me to be checked, it's just a model."

6 So, I mean, there has to be some understanding of
7 what a probabilistic inference means -- it doesn't mean
8 that you're more likely to be a bad guy or not, it just
9 happens to be this is the way we're checking. And I do
10 think that without transparency and people understanding
11 what being checked means, there's going to be a lot of
12 discomfort and anger.

13 So, we really do have to give people some
14 understanding of how the models work.

15 Ms. Toby Levin: Let me -- that's probably a
16 terrific example that bridges us into the next topic, I
17 want to get into a little more detail, on the redress
18 process. Because if people are -- if people's information
19 is being used in a data mining project, how do we address
20 redress when they may be very, either very removed, the
21 data collection was very removed from the activity that's
22 being -- it's underway. Or to use the example in a

1 screening program, where there may be a model -- maybe it's
2 random screening that's occurring, but people are being
3 caught up, and then there are a lot of assumptions made
4 that, you know, "What's wrong, why am I being singled out?"
5 And so, developing some understanding about a well-designed
6 redress program to help address that. But in a data mining
7 context, it's very difficult.

8 Does anyone want to take on the redress issue?

9 Mr. Fred H. Cate: Not really.

10 [Laughter.]

11 Mr. Fred H. Cate: But first of all, let me see,
12 I don't think we can in any way overstress how important
13 redress is. In part, for the reasons that Danny and I have
14 already talked about, that it serves at least two
15 functions. I mean, one is redress to the individual, and
16 one is feedback to the system.

17 But also because redress is an opportunity for
18 some of the transparency we've talked about before. You
19 know, one of the things I think we see from privacy in the
20 private sector, from privacy issues, is that the public
21 almost never uses the rights that they have, but it's the
22 knowledge that they have those rights that's very important

1 to building their comfort level or their trust.

2 We know this about privacy consent notices, we
3 know this about breach notices, we know this about all
4 sorts of opt-in and opt-out opportunities, that it hardly
5 matters what the mechanism is, they're not going to respond
6 to any of them, it's that they have had the chance to
7 respond.

8 So, one I think really clear lesson from this is
9 that no program ever should be rolled out without having an
10 incredibly visible redress mechanism, even if ultimately,
11 the redress -- because of classification issues, or because
12 of the data mining challenges -- has to be thinner than we
13 might like on this panel, you at least want to have it.
14 You know, you at least want to start with it as, you know,
15 that would be a leading point to have.

16 Second, and this goes back, partly to our
17 discussion about the classified information -- often it
18 seems when involved in redress in areas like airline
19 security, that there's a certain amount of unnecessarily
20 hiding the ball. That really, national security is not
21 going to be injured by the fact of giving me a general
22 description of where the data came from. Did it come from

1 the private sector or not? I mean, if that's going to key
2 the bad guys in, so that they can then get on planes in the
3 future, it's completely lost on me on how that is.

4 So that, the more that we can provide some
5 information, even if we can't provide all information, we
6 can avoid that sort of unfortunate circumstance of, "I
7 can't tell you. Sorry, too bad. Go away." You know, we
8 can engage in some sort of dialogue, even if the end result
9 is still, "I can't give you the specific piece of data you
10 would like to have."

11 So, again, it's one of the things I would say,
12 and maybe taking from the prior panel, we shouldn't let the
13 perfect be the enemy of the good; and, frankly, here, the
14 failure to do at least the good is one of the certain ways
15 to bring down even a fabulous system.

16 Mr. Thomas Oscherwitz: I think that's a really
17 critical point. And from -- as far as I can understand, I
18 think a lot of the -- I think there are a lot of untested
19 assumptions about what amount of redress is possible and
20 what isn't, given the constraints, you maintain a sense of
21 national security. And these are really technical
22 questions, these are, in a sense that these are data mining

1 questions, also, these are questions about information flow
2 -- how much information can you reveal to a particular
3 subject about the operation of a system, before you
4 compromise the whole system?

5 I think that unless there's a whole lot of
6 classified research I'm not aware of, I don't think this is
7 a question that's been studied much in its application to
8 redress systems, and it really ought to be. People who -- not
9 me -- but people who do cryptographic research, and do
10 research in information flow, can answer these questions in
11 relatively formal ways, and we could figure out what kind
12 of redress is actually possible, rather than just saying,
13 "Oh, no. If we tell you, we'd reveal X, Y and Z."

14 And so we should get serious about this -- this
15 is, you know, it's probably a little subspecialty of
16 various parts of information theory, but it's one that I
17 think really could be worked on quite profitably. And
18 rather than having this kind of knee-jerk reaction that
19 says, "Sorry, we can't tell you," there would be a more
20 systematic and trusted way of saying, "Well, this is what
21 we can tell you, and then there's stuff we can't." But we
22 ought to have a better basis for that.

1 Mr. Barry Steinhardt: I think the first thing
2 you said about redress is we need to have it. And it needs
3 to be real and it needs to be substantive.

4 The -- you know, we have of course the one
5 example, sort of glaring, staring us in the face, here,
6 which is the airline passenger system, where you have,
7 however many people there are on that list, whether it's
8 400,000 or a million, have lots of people who are impacted
9 by that. But there is no real redress, here, for someone
10 to get off that list, and whether they were put on the list
11 as a result of data mining or data matching, it's a little
12 hard to know.

13 One of the things that's missing that's essential
14 to any redress system, is an independent arbiter who's
15 making the final decision. You cannot have redress systems
16 that rely on the same people who made the operative
17 decision in the first place, doing the review and making
18 the final decision, you need to have an independent
19 official.

20 And we have to be careful about these claims,
21 that somehow, you know, if we have an independent official
22 that you know, the national security secrets are going to

1 be revealed, or that the bad guys are going to take
2 advantage of it, as if, you know, Osama bin Laden is going
3 to go into Federal Court in the United States, demanding to
4 have his name taken off the terrorist watch list -- those
5 things don't happen.

6 But, until there is some degree of genuine due
7 process, here, I don't think there's going to be a whole
8 lot of public confidence in the outcomes of these data
9 mining or data matching experiments.

10 Mr. Peter Swire: So, the question is, what about
11 redress, and the problem is that we might reveal our
12 secrets, and help the bad guys if we tell too much, that's
13 the way it's usually framed.

14 There's a saying in computer security circles
15 for, especially for the internet and open source, that
16 there's no security through obscurity. And this is, for
17 computer security folks, this is sort of a well-known
18 cliché -- there's no security through obscurity. The
19 thought is, if you publish your code, or show what crypto
20 system you're using, that you let everybody else sort of
21 whack away at it, you find the flaws, you fix the flaws,
22 you improve the system.

1 And there's a somewhat different saying in the
2 military side, which is "loose lips sink ships." If we
3 tell you anything, the U-boats will get the convoy and kill
4 everybody. And that's saying that secrecy's going to help
5 security.

6 So, I've written a couple of articles about when
7 each of those is correct -- when openness helps security
8 and when openness is going to hurt security, so to some
9 extent, you can go look at that. But there's a lot of
10 situations in a networked world, where people get to probe
11 a system over and over again, where if you think it's a
12 secret, you're wrong. Because the bad guys get to probe
13 over and over again, and they get to find out, and it's
14 often that you're keeping the secret from your fellow
15 defenders who aren't able to know what the flaws are, and
16 you don't end up fixing it.

17 So, the more that people -- if it's a distributed
18 system -- the more that there's lots of different people
19 probing it and attacking it, the more likely the bad guys
20 you care about already know what the flaws are, and so
21 you're sort of fooling yourself if you think secrecy is the
22 answer.

1 A second point is, that in defensive systems,
2 randomness is important. And so some fraction of the 2-
3 year-olds in the car seats and the grandparents should get
4 searched, because sometimes they'll be the ones carrying
5 the bad stuff, even though our data mining doesn't think
6 that's where the highest risk is. And if you have
7 randomness built in, occasionally people are going to get
8 searched and they're going to say, "Why did I get searched
9 on Thursday?" And the answer is, "We don't really have
10 much to say, except everyone's subject to search,
11 sometimes." You'd expect randomness to be part of the
12 answer, so you're not going to have full redress for the
13 occasional search, but there's a lot more room under the
14 sort of theory of when openness helps security, there's
15 probably a lot more room than we've seen to give redress
16 without revealing sources and methods, without compromising
17 the security that has been the instinct, I think, in the
18 last few years.

19 Ms. Toby Levin: We'd be remiss, I think, if we
20 didn't talk about one of the points on Fred's list, which
21 is the use of data minimization, and anonymization and
22 other tools -- basically, technology to help limit

1 information that's revealed or perhaps techniques that
2 might provide another approach to protecting privacy. Do
3 any of you have some thoughts on how we can better use
4 these tools? We had some introduction to it yesterday;
5 frankly it's very challenging for lay people to understand
6 these tools, so -- but I would, I definitely want to make
7 sure we've included that in our discussion this morning.

8 So, I'm going to -- I'm going to start with
9 Danny, who's been working with tools and technology for
10 decades.

11 Mr. Daniel Weitzner: Well, I might be the wrong
12 person to start with, because I have to say I come to this
13 with a certain amount of skepticism. I mean, these are --
14 in general, I think these are the -- these are a class of
15 tools, they're a set of cryptographic techniques that,
16 under controlled circumstances, allow you, for example, to
17 query databases and only get certain -- get answers back,
18 but only to a certain level of accuracy or answers about
19 the overall patterns in the database, without revealing
20 individual identity or without revealing information that
21 can be tagged to an individual person.

22 And these are exciting, they're sort of

1 tantalizing technologies, and they have, in particular
2 arenas, particular applications, and I'll let someone else
3 defend them. I'm going to tell you what I think their
4 limitations are.

5 I think that the limitation of these sorts of
6 tools is exactly what Peter said -- that it's hard to use
7 these tools in open, distributed, network environments.
8 When you can control, very tightly, the universe of people
9 who are asking a limited set of questions to a limited amount
10 of data, then you can have good privacy guarantees, but
11 when you have lots of people asking lots of questions, and
12 when those same people can go ask questions of other
13 databases that you don't control, then you have a very hard
14 time making any sort of guarantee about how much
15 information will leak out.

16 So, I think there's a place for these, but they
17 sort of travel under what I think are somewhat -- they're
18 sort of grand-sounding names like "privacy-preserving data
19 mining," and I frankly think that's not quite truth in
20 advertising.

21 So, I think that there is reason to apply these
22 where you can, but the labels may be a little bit

1 misleading.

2 Ms. Toby Levin: Let me ask you, then, in the
3 information-sharing environment where agencies are -- the
4 paradigm has changed from "need to know" to "need to
5 provide" and you have information sources across the
6 Federal government, many different databases, and agencies
7 would like to share information appropriately across these
8 databases; there may be some rules that apply to the data
9 coming from one agency, but does the other receiving
10 agency, then, have to respect those rules -- it gets quite
11 complicated.

12 Mr. Daniel Weitzner: Right, oh, and I hope
13 someone on the panel will defend these technologies,
14 because I could but I'm just not going to. I will if
15 pressed. But --

16 Mr. Peter Swire: Aren't you being a little broad
17 in your condemnation of privacy-protecting technologies,
18 here?

19 Mr. Daniel Weitzner: I'm not condemning privacy-
20 protecting technologies. I'm --

21 Mr. Peter Swire: Which large sub-set of them are
22 you condemning?

1 Ms. Toby Levin: Because you demo'd one.

2 Mr. Daniel Weitzner: No, I didn't use the term
3 "privacy-protecting," I used the term "privacy-preserving
4 data mining," which presumes that you can ask questions of
5 a set of data and only -- and be guaranteed that personal
6 information will not be revealed. And my -- it's a very --
7 you have to be very careful about what we're talking about
8 here, we're talking about this hope that, in particular,
9 before you get legal authorization to go do a subject-based
10 search, that you could go do a sort of a pattern analysis
11 of a bunch of data and get back something that's useful for
12 your investigation, but isn't privacy invasive -- that
13 doesn't uniquely identify anyone in that database. And my
14 assertion is that there are real limits to the ability to
15 make that guarantee.

16 It's a useful technology, particularly in --
17 there are circumstances in which it may be useful.

18 But, to Toby's question -- what do you do when
19 that doesn't work? So, that to me is the interesting
20 question. Sometimes, if these technologies work, then you
21 almost have no privacy problem, because you haven't
22 revealed any personal information, you haven't learned

1 anything about any individual, everyone's happy. What
2 happens when everyone isn't happy?

3 Well, then, I think you have to do just what you
4 suggest, Toby, you have to make sure that as information
5 moves around in a complex, analytic environment, that the
6 users of that data understand what the restrictions are on
7 how it can be used further on. That, if it came from the
8 airline passenger screening system, it better not end up in
9 the hands of the IRS for the purpose of catching someone
10 who is getting off an airplane, who is late on their taxes.
11 And that you do it, not by restricting the flow of the data,
12 but you do it by making sure that you understand the user
13 restrictions on that information as it goes through the
14 process.

15 Ms. Toby Levin: Let me ask just a clarification,
16 Danny. Is your point, though, about the technologies that
17 you can't ensure that they can't be -- that the identities
18 can't be revealed, by going back and re-identifying, or
19 somehow, actually figuring out who the individuals are?

20 Mr. Daniel Weitzner: Yes, yes. That you get a
21 little bit -- so, there are, you know, it's a very
22 carefully studied area of cryptography, and I will say, I'm

1 not an expert in this area, I just read the papers. But
2 the -- again, the guarantees about how much personal
3 information get revealed are very, very carefully scoped,
4 by people like Rebecca Wright who work in this area,
5 -

6 Ms. Toby Levin: But operationally, it would mean
7 that identities would not be part of the operational -- you
8 know, people using the data would not have those identities
9 revealed to them, not to say that they -- someone with
10 expertise could not go back in and deduce who that, the
11 identities of the individuals, but in fact, while this data
12 is being used, it would provide some obscurity.

13 Mr. Daniel Weitzner: Only if it's used in
14 isolation. And I just don't think investigations tend to
15 work that way. Investigators don't tend to ask one
16 question and say, "Oh, good. I now know that there are 17
17 people who are age 12 in this group, and I know I'm looking
18 at a 12-year-old, but I'm not going to look any further,
19 because I know there -- " the problem is, then I go and
20 say, "Well, what else can I find out about 12-year-olds,"
21 and then I find out are only five 12-year-old girls, I
22 don't know why I picked 12 -- 35 -- not to get into

1 conflict issues --

2 [Laughter.]

3 Mr. Daniel Weitzner: You know, I gradually piece
4 together information that -- where each piece of the
5 information is not directly attached to a person, but if I
6 ask enough questions of enough different sources, each of
7 those privacy-preserving answers will all of a sudden
8 become revealing, personally revealing, once I've asked
9 enough questions. And I think that's what investigators
10 tend to do, is they tend to want to -- they don't want to
11 just find interesting statistics, they want to find people.

12 Ms. Toby Levin: But from a security perspective,
13 the information would be less accessible.

14 Mr. Daniel Weitzner: On a moment-by-moment
15 basis, yes.

16 Mr. Barry Steinhardt: But that's not necessarily
17 -- that's not -- I think we're making the mistake here, of
18 you're using, you're referring to data minimization and
19 anonymization as examples of privacy-preserving
20 technologies, in which case I completely agree with you, I
21 think it's been dramatically oversold, and we should be
22 cautious about that.

1 That's not what I said, and that's not what this
2 Says. It says, using tools such as data minimization
3 and anonymization, so think about the airline security
4 example. We started out thinking we needed all of this
5 information from passengers to identify them. We wanted
6 their social, we wanted everything. Then, we later discovered
7 we only needed three elements to identify them. So, that
8 was data minimization, in action, it worked extremely well,
9 we can still identify the vast majority of passengers with
10 great ease, and now we're not burdening the system with a
11 bunch of other data. That's the type of use of this tool.

12 So, similarly, in the investigatory context, of
13 course, you want all the data, you want to get your hands
14 around it, and anything that denies you access to some
15 piece of relevant data, presumably raises -- or runs the
16 potential of raising a security issue -- but of course we
17 use data mining in all sorts of investigatory contexts.
18 Where we're not investigating something, we're just trying
19 to assess who should have access to this building, this
20 plane, this facility, what have you, and again there
21 anonymization, or tools to reduce who sees how much data --
22 even moving the screen that you see the backscatter x-ray

1 result, so that it's in a private place, rather than in the
2 middle of the airport terminal, I think these are extremely
3 important tools, and frankly, I think they're important not
4 just to the reality of privacy, but they're also important
5 to the public confidence, the public trust in the system.

6 Mr. Peter Swire: So, Danny was referring, I
7 think, to a literature about how hard it is to de-identify
8 someone, when you make queries into a closed database, and
9 you can also make queries into an open database. And it
10 turns out that a lot of the hopes for de-identification for
11 those people doing repeated queries, it's much harder to
12 de-identify than I used to dream that it would be. There's
13 a lot of results that show it's hard to stay de-identified,
14 when you can keep doing these queries. Is that roughly the
15 -- ?

16 Mr. Daniel Weitzner: Yes.

17 Mr. Peter Swire: Okay. So, that -- so, I agree
18 with Danny on that.

19 Let me be a little more optimistic, though,
20 picking up on Fred's approach on a variety of privacy-
21 related things you can do in the technology, and I have an
22 idea I'd never thought of before, I don't think it's been

1 proposed for -- putting on my old OMB hat.

2 I think that there might -- you know how there's
3 veteran's preferences for hiring in the Federal government?
4 What about privacy preference for procurement and research
5 contracts for the DHS? You know, you get some bonus --
6 maybe there would be a pot of money at S&T, maybe there'd
7 be a pot of money -- very large pot of money, right in the
8 Privacy Office, to do more workshops like this -- but also
9 a pot of money so that there's bonuses in the process
10 to encourage researchers and people proposing these systems
11 to have a clearly defined, privacy-enhancing aspects to
12 their proposals to the government. And if you had this pot
13 of money or Challenge Grants, or whatever, you need some
14 process to figure out who deserves it, and you'd have to
15 have criteria and avoid graft and corruption and all the
16 rest, but the point of having -- we have veteran's
17 preferences, why not have privacy preferences? Because
18 there are so many pressures the other way, that dangling some
19 dollars this way might have a good effect.

20 Mr. Thomas Oscherwitz: Maybe my vision isn't so
21 grand, but in the process, in building a data mining and
22 implementing a data mining model, there's a lot of ways

1 that you can minimize data access, whether it's the person
2 who's actually getting the results of the model, how the
3 people are managing it, how the result is actually -- is it
4 a score, is it a bunch of personally identifiable
5 information, there's a lot of ways that you can build
6 processes, where people work on it in an encrypted fashion
7 -- whether or not you can get to the perfect of having no
8 access, you certainly can minimize.

9 Mr. Daniel Weitzner: So that I don't end on a
10 pessimistic note, I want to --

11 [Laughter.]

12 Mr. Daniel Weitzner: I want to nominate my
13 current favorite privacy-preserving technology. I recently
14 heard about this. There is a hospital in Boston that was
15 one of the first hospitals to have a pretty comprehensive
16 electronic patient record system, so, you know, all the
17 tests, all the drugs, everything's all in there. And there
18 are terminals, you know, spaced around the hospital so that
19 everyone can get the data they need, this is great.

20 It happened that a very prominent doctor at the
21 hospital had a relationship with someone else in the
22 hospital, which ended in an unfortunate way, and it -- for

1 some reason, then, after that, that doctor was found
2 looking at this other person's medical records. And this
3 was a publicly reported case, the doctor was actually --
4 probably should have been fired, but he was demoted,
5 substantially, from his august position for doing this.

6 And the way that he was caught was really
7 interesting, and this is my nomination for privacy-
8 preserving technologies. It happens that the way this
9 system was designed -- this is an old system, it's not a
10 new system -- this system is designed in such a way that
11 when you pull up a patient record, what you see on it, on
12 the first screen, is you see a list of the last 10 people
13 who accessed that record. That's it. It doesn't tell you
14 whether they were -- it doesn't tell you why they
15 -- why they accessed it, it doesn't do anything fancy at
16 all. It just says, here are the last 10 people.

17 So, this woman's doctor, when he looked at her
18 record -- this is her actual doctor, not the doctor with
19 whom she had the relationship -- her actual doctor saw this
20 record and said, "Well, why did that guy look at her
21 record? He's not her doctor, he doesn't treat her," you
22 know, maybe he knew some other things, and so this is just

1 a really, really simple little piece of technology again,
2 that sort of makes information usage more visible, and
3 allows for the application of human intelligence and
4 integrity, and really caught what was a very serious
5 violation of privacy rules.

6 Ms. Toby Levin: Well, that's a good segue to my
7 last question which -- that type of log-in control is
8 something that we look for very frequently in our -- when
9 we do our Privacy Impact Assessments, which are the bread
10 and butter operation of the Privacy Office, in terms of
11 making sure that privacy is part of the activities at the
12 Department.

13 So, Peter, I think, made a reference to PIAs, and
14 possibly thinking in terms of additional questions or
15 elements that could be added to PIAs for, with regard to
16 data mining, and so I'd like to sort of close our panel
17 today with just -- if there are any other suggestions with
18 respect to the PIA content or process that you think would
19 be helpful, in light of the issues that we've been
20 discussing.

21 Mr. Peter Swire: One part -- this is a longer
22 conversation is, there's a Civil Liberties Office at DHS, and

1 there's civil liberties issues about, that have to do with
2 sometimes with effects on different ethnic groups or a
3 variety of other things.

4 It seems to me, at least worth considering, whether
5 Privacy Impact Assessments should be combined with any
6 civil liberties assessments.

7 A second point is, now that we've had -- the eGov
8 Act of 2002 said these Privacy Impact Assessments should
9 happen for new systems. I don't think they yet apply
10 routinely -- certainly not at the Federal level, to
11 regulations, and I think that they should, and so there
12 might be some wider range of things that should trigger the
13 PIAs.

14 A third point is, if you have a PIA, maybe with
15 supplementary questions for data mining, there might be
16 other categories you get experience with, for instance, I
17 think biometric systems would be a good candidate for
18 having supplementary questions for PIAs. And as you
19 develop experience in various sectors, as the Privacy
20 Office, developing supplementary things that don't apply to
21 everybody, but apply when there are certain areas that have
22 certain kinds of issues that recur, I think that that can

1 be sort of a next generation, that maybe you're on the path
2 for already. I don't know all of the details, but I think that
3 could deepen what we got from 2002.

4 Ms. Toby Levin: I'd like to thank the panel, and
5 we'll have some closing remarks, so thanks very much.

6 [Applause.]

7 Ms. Toby Levin: Oh, I'm sorry, I forgot our
8 question period, I'm jumping the gun.

9 Okay, so if we have some -- anyone who would like
10 to pose a question?

11 Ms. Jennifer Schiller: Hello, Jennifer Schiller
12 from S&T. I have about 50, but I'll limit myself to 2,
13 because I'm sure there are others who have them.

14 Toby had asked what you would say to the
15 researchers to convince them to implement these privacy
16 processes, and I would just say, our researchers are
17 subjected to a variety of mixed messages. From Congress,
18 we get, "Focus on domestic radicalization, focus on
19 suspicious behavior detection, but don't use any personally
20 identifiable information." We get, "Be fast, be nimble, be
21 effective, but jump through 5,000 bureaucratic hoops
22 between your idea and an implementable technology."

1 So, I would ask, my first question is what would
2 you say to our researchers and to our Privacy Office, to
3 help us work together to navigate that tension?

4 And my second question is for Professor Cate --
5 you had talked about standards. And I have a question
6 about standards when it comes to the sources of
7 information. Is information that's Google-able, for lack
8 of a better word, you know, we create research databases,
9 for instance, newspaper articles about terrorist events,
10 that might have the name of like, a U.S. citizen in it.
11 When we create those types of collections of information,
12 is there a lower standard of privacy protection, given that
13 this information is in the public arena, these are public
14 figures, public events?

15 So, these are my questions.

16 Mr. Fred H. Cate: May I start with the second
17 question?

18 I think the answer is yes, although I wouldn't
19 phrase it that way. In other words, I wouldn't say there's
20 a lower standard, I would say the standard is more easily
21 met. And so, if it's publicly available information,
22 obviously for efficacy reasons, you still care about the

1 provenance of the information, is it useful, is it
2 accurate, is it going to lead you to an outcome that
3 actually enhances security?

4 I wouldn't, in that sense, think of lowering any
5 of those standards. But, I think in terms of thinking
6 about the privacy impact on the individual of accessing or
7 using that information is much easier -- it's going to be
8 much easier to satisfy whoever's going to authorize that
9 program, that, "Well, this is widely available, it's
10 publicly accessible, it's accessible without charge," that
11 would diminish whatever sort of privacy interests we might
12 think -- and hear, in that information.

13 If I could just also add, with response to your
14 first question, first of all, it doesn't help a bit, but I
15 could not be more sympathetic. I think the lack of
16 consistency and clarity -- from Congress, in particular,
17 here -- has made everybody's job in the national security
18 arena, much more difficult.

19 But, having said that, I also think that there
20 are things that can be done to, you know, you've got to
21 operate, even with an inconsistent, somewhat ineffective
22 Congress. And I don't live in a world of sort of enhanced

1 bureaucracy. In other words, I'm not looking to give you
2 five new forms to fill out before the researchers can get
3 on with their job.

4 I, in particular, think of privacy and security -
5 - especially in the research environment -- as almost
6 always intrinsically linked. That you're going to get a
7 better outcome in terms of detecting terrorists, if you are
8 also privacy sensitive. Because the same types of
9 questions -- the accuracy of the data, the efficacy of the
10 system, the targetedness of the inquiry -- all of those, I
11 think, are going to help in both.

12 So I would hope, and part of what we're doing is
13 accentuating steps that you would think the researchers are
14 already going through, but maybe haven't thought of in
15 these privacy-enhancing terms. And that is going to
16 be a practical step for the researchers and the Privacy
17 Office in the future, to sort of re-characterize some of
18 what's already going on, without just adding levels of
19 bureaucracy, to add more.

20 Ms. Toby Levin: Yes, go ahead.

21 Chris?

22 Mr. Chris Clifton: Believe it or not, I'm not

1 here to debate Danny about privacy-preserving data mining -

2 -

3 Ms. Toby Levin: Right, this is Chris Clifton,
4 for purposes of the transcript.

5 Mr. Chris Clifton: That's my hobby horse, but I'm
6 here to go after the whole panel, actually. Fred made a
7 point, very briefly, about a second advantage to redress
8 mechanisms, and that is feedback. And then there was a lot
9 of discussion about, well, feedback, or redress mechanisms
10 help you build trust, and it's not even important that they
11 be used. I think the Privacy Office really needs to sell
12 redress mechanisms as a feedback mechanism -- this is the
13 way you make your systems better, is you find out where
14 you're failing, and this is how you do it. And make it
15 easy for people to provide that feedback.

16 Ms. Toby Levin: I would like to say that our
17 email address is probably used by more members of the
18 public than any other email address at the Department. We
19 do have a very public email address, privacy@dhs.gov, so
20 we, I'm sure we'll be hearing more about that, but I
21 certainly agree with your recommendation.

22 Next?

1 Mr. Daniel Weitzner: Could I just quickly
2 respond to Chris -- I think that, I think that is an
3 important lesson from the private sector, that's an
4 important lesson from credit reporting systems, that what
5 we know is that people pay attention to data quality when
6 it hurts them.

7 What we also know, as Fred was saying, people pay
8 no attention to privacy or data quality when they're just
9 being asked for their name for the 15th time that day, and
10 they need to give it anyway to get the book they want from
11 the website, or whatever it is.

12 And so, I think really looking closely at how
13 those kind of large-scale systems work and maintain,
14 obviously, not perfect data quality, but do better with the
15 interaction, I think, is quite valuable.

16 Mr. Fred H. Cate: Just to echo -- and that's not
17 happenstance. In other words, that's a calculated -- you
18 know, Congress made a choice back in 1971 that credit
19 information furnishers only had to make efforts to be
20 reasonably accurate. They would get the rest of the
21 accuracy out of letting, if you will, victims of inaccuracy
22 get free credit reports, get a chance to respond.

1

2 Over the past 30 years, we've seen more and more
3 rights added on, you know, mandatory dispute resolution,
4 opportunity to correct -- that's not a bad model, frankly,
5 here, as well. To say, you know, you can only do so much
6 up front, the cost of making it perfectly accurate would
7 sink the ship, but that only works if you then have a
8 meaningful redress system to come back around, to make it
9 more accurate as you discover inaccuracies.

10 Ms. Toby Levin: Okay, next question, please?

11 Ms. Jennifer O'Connor: I'm Jennifer O'Connor,
12 S&T, and I'm one of the research program managers that
13 Jennifer keeps referring to, that's had problems with this
14 whole issue. And I'll just kind of make you aware, from our
15 perspective, of some of the issues that have come along.

16 The issue of, when do you classify something -- I
17 have a program, the program I'm managing, that deals quite
18 a bit with social-behavioral science. We have actually
19 stepped back -- but it's using data mining tools, depending
20 upon how you define data mining in terms of modeling
21 techniques --

22 Ms. Toby Levin: We're having a little trouble

1 hearing you, could you raise the mic, please?

2 Ms. Jennifer O'Connor: We're actually using data
3 mining tools, but we're using social and behavioral
4 sciences. And we're actually doing something that most of
5 the data mining folks aren't doing. We're actually
6 taking the social science and trying to show how you can
7 take theories from a semantic standpoint, you know, based in
8 social science, and apply them to why you
9 would use certain variables in a data mining enterprise.
10 Which, those are two -- that's actual research. I mean,
11 for those of you, I mean, you guys are familiar with data
12 mining and a lot of cases, when you're training an
13 algorithm, there's no theory there, you're just letting
14 whatever comes out come out.

15 What we ran into, though, unfortunately, was when
16 we went to the Privacy Office with this, we didn't have
17 answers, we had a bunch of theories in terms of what we
18 wanted to test. A year and a half later, after our PTA
19 went through -- meanwhile, I'm trying to keep this
20 funding alive, which our division doesn't have very much of
21 to begin with, because it's social-behavioral science,
22 we're not predicting -- you know, we're not protecting

1 critical infrastructure, we're not stopping people at the
2 border, we're trying to stop radicalization, basically.

3 My question to you would be, a) when you're
4 dealing specifically with social-behavioral scientists,
5 unfortunately anybody that works in government seems, for
6 some reason, to think they are a social scientist of some
7 sort --

8 [Laughter.]

9 Ms. Jennifer O'Connor: We get asked these
10 questions that it would be great to be able to take to an
11 IRB, instead of going back and forth, and I -- Jennifer you
12 can correct me -- we went back and forth probably 6 times
13 with your Privacy Office clarifying scientific principles
14 in social science that I guarantee you wouldn't have asked
15 of somebody doing critical infrastructure or specific data
16 mining techniques, or et cetera. And there's got to be some
17 balance there --

18 Ms. Toby Levin: Well, let me explain, though,
19 part of that is because most PIAs are public
20 documents, and they need to be written in a way that the
21 public can understand them, so we look at it, "Can we
22 understand the principles? Can we understand the responses

1 to questions?" If we can't, then how can the public
2 understand? So, part of it is that.

3 Ms. Jennifer O'Connor: Then explain the medical
4 community, because the medical community -- if I was a
5 doctor trying to explain why Chemical X and Chemical Y work
6 together to do that, they have a real simple answer, "It
7 stops this." We should be able to give you the same
8 answer. It stops this.

9 But, we're not. You guys dig down further, and
10 unfortunately some of what we have is fragile data, the
11 stuff that really works, yeah, a law protects it. You
12 know, you cannot stop giving that information because the
13 law says, "You will get this information as a United States
14 citizen." There's some stuff that, we don't know whether
15 it works yet, or not, that's fragile. If it does work, and
16 it gets out in the public information through a PIA, for
17 whatever reason, it's dead, it no longer works. The whole
18 research effort is no longer useful.

19 And that's the classified, not because -- I have
20 worked very hard to keep my program unclassified, because I
21 strongly believe that people have the right to know what
22 we're doing. And it would be -- it would have been a lot

1 easier if I went into the classified realm, and didn't have
2 to deal with some of these issues.

3 And I guess, there needs to be an education,
4 there needs to be an education, both in terms of the
5 Privacy Office's need to understand that they don't
6 necessarily need to understand -- specifically in the
7 social sciences, and some of the human factors areas --
8 exactly what's going on, because there is science there.
9 There needs to be some trust from the Privacy Office's
10 standpoint that when we tell you we're using social science
11 to actually data model, that that is science.

12 Ms. Toby Levin: Right.

13 Ms. Jennifer O'Connor: So, I guess --

14 Ms. Toby Levin: Well, we have a complex role,
15 because we're also acting as, in a sense, as an auditor at
16 the early stages of scientific research. So, I guess the
17 question is, when you've got projects at very early stages,
18 as she described, what are the ways that the privacy issues
19 can be best addressed?

20 Ms. Jennifer O'Connor: Like an IRB, when -- I
21 mean, can the Privacy Office have an IRB on call, that
22 actually is an expert that we can go to and talk to, maybe

1 even at the classified level?

2 Ms. Toby Levin: All right, Danny? Thanks.

3 Mr. Daniel Weitzner: I have to be a little
4 pessimistic here again, for a moment, Jennifer. There's an
5 interesting divide as to IRBs in the academic community.
6 The medical community may argue about whether they're
7 effective or not or what they're -- but they're sort of a
8 well-worked out process.

9 Most people in social sciences, at least, that I
10 know, hate IRBs. And that's because to a very large
11 extent, you have social scientists who are often studying
12 what is, essentially, public behavior, they want to look at
13 people who are just out there in the world. And, you know,
14 you look at IRBs for medical research -- you know, they
15 have to make sure that people are not going to get hurt.
16 And, where hurt is a relatively well-understood quantity.

17 When you start applying IRB mechanisms
18 to social science -- and I fear this would apply in a lot
19 of cases in which personal information is involved that's
20 not medical -- you have a hard time putting a, sort of
21 getting your head around what the mission of the IRB is
22 supposed to be.

1 If it's supposed to enforce the privacy rules,
2 and we've already said, "Well, it's a little, you know,
3 we're trying to establish what those are," what is it that
4 we're going to tell the IRB to do?

5 So, I think that it's quite reasonable to expect
6 an expert group to be able to look at a research program.
7 To expect that the group that's evaluating the research has
8 expertise in the area. That makes a lot of sense. I
9 wouldn't want to get too enthusiastic about just, throw
10 some IRBs in there, because there's the beginning of, I
11 think, a real rebellion, actually, amongst social
12 scientists -- particularly anthropologists and sociologists
13 -- in having their research subject to IRBs, at all, so --

14 Ms. Toby Levin: So, the question is, how can we
15 respond to help in this situation?

16 Mr. Fred H. Cate: But there is one, you know,
17 silver lining to your dark cloud there, Danny --

18 Mr. Daniel Weitzner: I'm usually a pretty
19 optimistic person, I don't know what's wrong with me.

20 [Laughter.]

21 Mr. Fred H. Cate: And that is, places where IRBs
22 have worked better for social science researchers, it's

1 because it reflects a real expertise of the membership of
2 the IRB, including social scientists.

3 And so one of the things that is clever in the
4 theory -- and occasionally in the practice of IRBs -- is
5 usually they are composed of a pretty broad variety of
6 members, who bring expertise in different areas of the
7 types of proposals they'll be looking at, the theory being
8 that that way there is always somebody who can do the sort
9 of translation thing, that, you know, bring the context of
10 the background of research norms from that field. And then
11 the body, as a whole, can help do the -- how do, do we need
12 to explain this differently, do we need to think about a
13 mechanism for protecting the human interest here? That
14 aspect, I think, would still be quite useful, even if not a
15 formally structured IRB.

16 Ms. Toby Levin: And maybe Privacy Office staff
17 will be able to sit down with you and learn a little bit
18 more after this session, as to how we can better serve --
19 or best serve -- your operations and understanding how to
20 do these PIAs better. Happy to commit to doing that.

21 Michael?

22 Mr. Michael A. Aisenberg: Three comments from

1 the -- hopefully private sector perspective on some things
2 that were just said. The first will be on Peter's
3 incentives and preferences point, the second on a possible
4 tool, based on aging of the data and the third on, just an
5 observation about public records.

6 There has been proposed, a preference for
7 industry, in contracting with the Department and other
8 agencies of government, for industry members from the
9 critical infrastructures of the industry ISACs. So -- the
10 Information Sharing and Analysis Centers across the various
11 critical infrastructures, so the notion of preferences for
12 those who can demonstrate a -- some sort of a privacy-
13 conforming practice, as well, is not off the reservation
14 very far, and I think it makes a lot of sense.

15 The second point is, in terms of industry data
16 retention practices right now -- even though the cost of
17 storage has crashed dramatically over the last generation,
18 storage is still expensive, and secure storage -- reliably
19 secure storage -- is expensive, and so a lot of people age
20 their data, or obliterate data, or let caches expire
21 through TTL, Time to Live tools that wipe out data at a
22 certain point. And I think for certain classes of PII that

1 are being collected, meta-tagging data with a time-stamp of
2 its date of origin, and a time to live that would cause
3 that data to blow up at a certain point, where the argument
4 for its use has evaporated could readily make a lot of
5 sense.

6 The third point is simply one about public record
7 compilations. So, we had an experience in the President's
8 NSTAC a number of years ago -- this is very similar to the
9 project that came out of the Geographer from George Mason
10 University -- a consultant to the President's NSTAC was
11 asked to compile a list of the 20 most important
12 telecommunications and 20 most important information
13 technology physical installations. That research project
14 was conducted entirely from zoning records and public --
15 other public documents. The compilation -- that list,
16 along with Google maps, in full color, showing exactly
17 where these buildings were and how to get to them -- were
18 put together in a report that was left on the table of
19 every member of a classified session of the President's
20 NSTAC.

21 Needless to say, those reports were picked up,
22 individually and burnt afterwards, because the compilation

1 had turned into a very valuable, and necessarily
2 classifiable tool. So, the same thing, in terms of -- this
3 goes back to the point I was making about the different
4 faces of the Rubik's cube -- one party's data mining
5 exercise can become another party's act of war, and I think
6 we have to be sensitive to the fact that the sources of
7 that data can be very benign public records, can be, for
8 example, as simple as a telephone book.

9 Ms. Toby Levin: Okay, thank you.

10 Next?

11 Mr. Andrew Feinberg: Hi, Andrew Feinberg from
12 Washington Internet Daily.

13 I'm --

14 Ms. Toby Levin: I'm sorry, I couldn't hear you?

15 Mr. Andrew Feinberg: Hi, Andrew Feinberg from
16 Washington Internet Daily.

17 I wanted to respond to -- or, let me ask a
18 follow-up to the previous question on Google-able data.
19 And I believe Mr. Cate had said that there could be a lower
20 standard for using publicly available data in that sort of
21 sense, but when you're looking at records collected by
22 private companies, like Google, a lot of these records,

1 people aren't aware that they're being made public, or that
2 they're even available to these spiders, these crawlers
3 that gather the information.

4 So, how do you make that distinction? Because
5 when I give my records to a credit agency or to someone who
6 is checking, you know, for an apartment or for anything,
7 that's -- I know I'm giving it to somebody. I don't know
8 everything that is in that cloud of data that Google and
9 other companies like that have on me. So, how can you --
10 how do you reconcile that?

11 Ms. Toby Levin: I think we keep coming back to
12 this question of, you know, of how we deal with the fact
13 that so much information is being collected and used in
14 ways that the public just has no awareness.

15 Mr. Fred H. Cate: Well, first of all, I just
16 want to be clear, I don't think I said a lower standard, I
17 think I said the standard would be easier to meet, if it
18 were publicly available data.

19 And, also the question had to do with Google-able
20 data, not data held by Google. So, there's a huge
21 difference there. In other words, if the data is out there
22 where anybody can get it, at any time, for free -- so, the

1 phone book, the internet, places like that, I think it's
2 harder to make a really strong privacy argument surrounding
3 that data.

4 If we take the data that Google is collecting
5 about my searches, which is not accessible to anybody
6 anytime for free, I wouldn't make that argument at all. In
7 fact, I would make the opposite argument, that this is data
8 that is not disclosed to the government, this is data that
9 the government would be getting from the private sector,
10 it's not data that's publicly available, and so we would
11 expect the privacy standards to be even harder to meet.

12 And one of the things that pretty much all of the
13 prior groups to look at these issues, and I think, most of
14 the conversation on the past two days has suggested, is
15 that where data is completely repurposed, so for example,
16 it moves from the private sector where it was being used to
17 complete a transaction, to the public sector where it's
18 being used for some other purpose, the privacy issues
19 become, frankly, at their most acute, and -- for lots of
20 reasons. Because, we have all sorts of questions about the
21 accuracy of the data for this new purpose, because of the
22 sort of the transparency issues, because of the privacy

1 impact issues.

2 So, again, it's not at all to be nit-picky, it's
3 to say that the terms make a huge difference here. So,
4 really, publicly available data, to me, would be at one end
5 of the spectrum of, say, concern, and data held by private
6 industry -- I don't mean to point to you, Tom -- and that
7 is not publicly available, and is not normally held by the
8 government, would be at the far other end of the spectrum
9 of concern, in my way of thinking.

10 Mr. Andrew Feinberg: Can I follow-up, though?
11 When you said "publicly available," and she said "Google-
12 able," to me that means almost the same thing. I know what
13 you mean by publicly available, but to me, if this private
14 company, using some kind of algorithm that we don't know
15 what they're using can find it somehow, and therefore, it's
16 out in the public sphere. And associated with, let's say,
17 my name, someone just searched for my name and this
18 document comes up. Obviously, that is publicly available;
19 we don't know how it was made publicly available, though.

20 Ms. Toby Levin: Let me just interject a
21 statement with -- in defense of the Privacy Act. I know
22 there's been points made today about its limitations, and

1 questioning whether it's up to the challenge of the
2 information world that we're living in today, but -- if
3 information, even if it's publicly available -- is brought
4 into a Department of Homeland Security or other Federal
5 agency system of record, into a database where it's used as
6 part of our government operations, that information --
7 regardless of the fact that it came from a Choice Point, or
8 Axiom, or an information broker, or came from a public
9 website -- if that information is brought into the Federal
10 government, into a database, into a system -- that
11 information then becomes subject to the Privacy Act, and
12 the protections that it affords.

13 So, I think there's sometimes confusion about
14 that, so once we bring it into the government operations
15 and into our systems, and we're actually making it part of
16 our operations, that is subject to the Privacy Act.

17 Mr. Andrew Feinberg: And that's any information?

18 Ms. Toby Levin: Excuse me?

19 Mr. Andrew Feinberg: That's anything?

20 Ms. Toby Levin: Right, it doesn't matter what
21 the source is of the information. The fact that it's
22 brought under our umbrella, yes.

1 Mr. Andrew Feinberg: Thank you.

2 Mr. Thomas Oscherwitz: Just a brief comment, I
3 think that you do raise a good point, which is that I think
4 there's an incredible trend right now that more and more
5 personal information is going into the "cloud" which is
6 under -- not necessarily under control by, you know, any
7 party, but it's just out there. And I think we're going to
8 see that going on more and more, and I think that has to be
9 addressed.

10 Ms. Toby Levin: Barry?

11 Mr. Barry Steinhardt: Yeah, let me just make
12 several disconnected points, here, to respond to various
13 things that have been suggested.

14 One is, that my understanding of the Privacy Act,
15 Toby, is that it -- the information that comes into the
16 Department of Homeland Security or any other Federal agency
17 is not subject to the Privacy Act unless it's as part of a
18 system of records.

19 Ms. Toby Levin: I don't understand. You're
20 saying it's not part --

21 Mr. Barry Steinhardt: I'm sorry?

22 Ms. Toby Levin: You're saying it's not part of

1 the -- ?

2 Mr. Barry Steinhardt: That, the kind of notice
3 requirements that are attendant to the Privacy Act do not
4 kick in until you do something beyond just collect a single
5 piece of data, which is what you're suggesting, here.

6 Ms. Toby Levin: Well, the ad hoc, there's --
7 The ad hoc collection of pieces of information has been
8 interpreted to be outside of the Privacy Act.

9 Mr. Barry Steinhardt: Right.

10 Ms. Toby Levin: But, when we're systematically
11 going out and collecting information and putting it into
12 our records, that are used either for transportation, or
13 for border protection, or any of our systems - if that's
14 systematically the way we operate, then that would be
15 included.

16 Mr. Barry Steinhardt: I think we're saying the
17 same thing.

18 The other point I wanted to make, though, was
19 about Google. You know, Google is also sort of cited as
20 the reason why we have no more privacy.

21 [Laughter.]

22 Mr. Barry Steinhardt: Poor Google. And, but you

1 know, I think that that's an exaggeration, at this point,
2 for a couple of reasons. A lot of the most sensitive data
3 that people are concerned about -- their medical records,
4 their financial records, are not available -- at least
5 until Google takes over the Medical Information Bureau --
6 but, you know, they're not available right now by Googling
7 -- there may be little episodic portions of it, you know,
8 you may be able to go and determine that somebody
9 subscribes to a cancer awareness bulletin board or
10 something -- but overall, that information is not yet
11 available for that kind of public searching.

12 And the second thing is the notion that somehow
13 Google, and some of these similar enterprises, are somehow
14 exempt from, you know, the rule of law. We had a really
15 interesting development recently, involving Google Earth,
16 not Google Earth, I'm sorry, Google Street View, that has
17 these very detailed photographs, right? You know, of
18 street views that show personally identifiable people,
19 show license plates, other personally identifiable
20 information.

21 There was a discussion about the possibility of
22 that technology being introduced into Canada. And the

1 Canadian Privacy Commission said -- whether it was going to
2 be or not is another matter -- but the Canadian Privacy
3 Commission said to Google, "Well, you know, we have some
4 laws here about personally identifiable information, and
5 you just can't come into Canada and offer to everybody a
6 view of a street in Montreal that reveals this personally
7 identifiable information." To which Google -- they've had
8 the exact same experience in Europe, more recently -- to
9 which Google's response was, "Okay," eventually, "We
10 understand that. Here's what we're going to do. We are
11 going to blur -- or use the available technology -- to blur
12 the images so that they're no longer personally
13 identifiable."

14 Now, is that a sort of perfect solution? Can you
15 get around that? No. But it's a pretty good solution.
16 It was sort of -- I took that as an important object
17 lesson which is that, if you have laws that require that
18 privacy be protected, that the technologists will find the
19 technical solutions to accomplish that. And that's largely
20 what we are lacking in the U.S., is that we don't have
21 those laws that require all of these technologies that
22 we're talking about be in place, at least with respect to

1 the Federal government.

2 Mr. Andrew Feinberg: That was my next question,
3 if you thought that that would bring a need for the U.S. to
4 have similar privacy legislation, but it seems you answered
5 it.

6 Mr. Barry Steinhardt: My answer is -- I'm sorry,
7 I'm having a little trouble hearing.

8 Mr. Andrew Feinberg: I think you answered my
9 next question, which was going to be if you thought the
10 U.S. needed similar privacy legislation.

11 Mr. Barry Steinhardt: Yes, I think the U.S. has
12 to get with the rest of the developed world.

13 [Laughter.]

14 Mr. Barry Steinhardt: And both have overarching
15 privacy laws, and independent privacy officials. It would
16 be nice if we joined the 27 or the 29 OECD nations that
17 have those rules.

18 Ms. Toby Levin: Okay.

19 Mr. Andrew Feinberg: Thank you.

20 Ms. Toby Levin: That will be the last question -
21 - thanks very much.

22 Well, I hope all of you feel as optimistic as I

1 do now that we've set things in motion. I think we've --
2 the Privacy Office is very appreciative of your
3 participation and the cooperation that we've had with the
4 Science and Technology Directorate to move forward in
5 actually coming up with principles that can provide a
6 roadmap for the Department, so that it can do what it's
7 really authorized to do, which is to do data mining to help
8 protect the homeland, and which is also -- which it should
9 do in a way that protects and preserves privacy, because
10 that's, I think, ultimately what we all want when we say
11 "preserve the homeland," it is a homeland that is
12 respectful of privacy.

13 We will extend the period for written comments to
14 August 15th, so hopefully if any of you have additional
15 information you'd like for us to consider, you'll submit
16 it.

17 I particularly want to, again, thank our team
18 that helped pull this workshop together -- Sandra Debnam,
19 Sandra Hawkins, Tamara Baker, and Rachel Drucker, who were
20 specifically very helpful in supporting our work here.

21 And we hope you'll come again to these workshops,
22 I think they're informative for all of us as policymakers,

1 and as people with an interest in these highly -- sometimes
2 controversial -- but important issues, and we look forward
3 to hearing from you in the future, by whatever vehicle --
4 whether it be email, or by your attendance.

5 So, thank you, again.

6 [Applause.]

7 [Whereupon, at 12:43 p.m., the meeting was
8 adjourned.]

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