



U.S. PRIVACY AND CIVIL LIBERTIES OVERSIGHT BOARD
(PCLOB)

PUBLIC FORUM ON ARTIFICIAL INTELLIGENCE

Thursday, July 11, 2024

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A P P E A R A N C E S

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P R O C E E D I N G S

MS. FRANKLIN: Hello. I'm Sharon Bradford Franklin, chair of the Privacy and Civil Liberties Oversight Board. Together with my fellow board members, Ed Felten, Travis LeBlanc, and Beth Williams, I'd like to welcome you to today's public forum on the role of artificial intelligence and counterterrorism and related national security programs and the privacy and civil liberties issues associated with these uses of AI.

The uses of AI in all facets of our lives are rapidly and continually growing, as is the sophistication of these tools. These trends are raising a variety of questions for policymakers, ranging from overarching concerns like how to encourage American competitiveness in AI and what limit should be put on AI to avoid harmful outcomes. To more day-to-day and specific questions, such as how to prevent cheating by students who want to use ChatGPT to write their research papers.

But the Board's focus on AI is not motivated

1 simply by a desire to get in on the latest big tech
2 graves. Rather, as our government incorporates AI
3 tools into its efforts to protect the nation from
4 terrorism, it is our role to ensure that those
5 government strategies also protect individual rights
6 and liberties. Yet the potential uses of AI are
7 extensive, and we must be strategic in our oversight
8 of the government's use of AI for counterterrorism
9 purposes.

10 Even before most people became aware of
11 generative AI and tools like ChatGPT, there's been a
12 lot of research on the privacy and civil liberties
13 risks posed by AI tools. As I expect, we will discuss
14 further today, these range from reliance on training
15 data that reflects and perpetuates patterns of
16 historical discrimination, to bias and facial
17 recognition systems that don't work as well for
18 particular racial or other demographic groups, to AI-
19 based decision making that lacks explainability or
20 other due process safeguards.

21 On the other hand, AI offers enormous
22 benefits. And I'm encouraged by the various efforts

1 to develop frameworks to address the risks posed by
2 the uses of AI. In recent years, these have included
3 the artificial intelligence ethics framework for the
4 Intelligence Community, and the White House blueprint
5 for an AI Bill of Rights.

6 Just last fall, the President issued a new
7 executive order on the safe, secure and trustworthy
8 development and use of artificial intelligence. And
9 we expect the issuance of a National Security
10 Memorandum, or NSM, on AI before the end of this
11 month.

12 But how well is the government doing it
13 translating these principles into action? What gaps
14 remain in these frameworks? And how can the Board
15 best focus our resources to conduct our oversight on
16 the use of AI in counterterrorism?

17 So, today's public forum is designed to
18 inform both our Board and the public as we work to
19 scope and define our oversight of the government's use
20 of AI in counterterrorism and related national
21 security programs.

22 Before we turn to that conversation, I have

1 just a couple of notes on today's event. First, I
2 want to thank our staff for all their tremendous work
3 planning today's forum and making it possible for us
4 to come together online. And then in terms of
5 logistics. Today's forum of course is limited to
6 matters that can be discussed in this unclassified
7 public setting. Our format will include remarks from
8 Senator Rounds, followed by two panels.

9 Today's event is being recorded, and the
10 recording will be posted on our website. For each
11 panel we will first hear brief opening statements from
12 each panelist, then my fellow Board members and I will
13 take turns asking questions of the panelists. And we
14 will cycle through our order again as time permits.

15 So, we will begin our event today with some
16 pre-recorded remarks from Senator Mike Rounds. He is
17 co-chair of the Senate AI Caucus and a member of the
18 bipartisan Senate AI Working Group. And he also sits
19 on the Senate Select Committee on Intelligence.

20 So, over now to those remarks.

21 UNIDENTIFIED SPEAKER: My apologies. I will
22 start video with audio.

1 MR. ROUNDS: Hi, I'm Senator Mike Rounds.
2 Thank you for the opportunity to talk to you today. I
3 know that the use of AI by the Intelligence Community
4 for counterterrorism and other IC missions is a topic
5 of particular interest to you.

6 Our Intelligence Community collects enormous
7 amounts of multi-source data each day that the United
8 States uses to support national security priorities
9 and objectives to include counterterrorism. AI is
10 capable of processing huge amounts of data, which in
11 turn is being utilized to identify patterns of life
12 and to analyze significant amounts of data in a very
13 short period of time. Such identification and
14 analysis can continue to enable the IC to more
15 effectively and efficiently track suspected or known
16 terrorists as well as terrorist financing activities.

17 For example, the IC's Project Maven uses an
18 AI tool designed to process imagery and full motion
19 video from unmanned systems and can systematically
20 detect potential targets for collection. This will
21 allow us to more efficiently identify and neutralize
22 terrorists.

1 As a member of the Senate Select Committee on
2 Intelligence, I look forward to further fostering
3 these efforts. As our government matures these
4 capabilities, it will be important to establish
5 metrics to measure the performance and efficacy of
6 these AI supported capabilities. Such performance
7 metrics for the IC's use of AI could include measuring
8 the speed of analyzing intelligence data sets across
9 collection platforms and the breadth of resource
10 utilization, as well as the depth of global
11 collection, all balanced against protecting liberties
12 and Fourth Amendment privacy protections.

13 I believe that measuring these kinds of
14 performance metrics must be a part of the intelligence
15 committee's oversight of the IC. As we foster new IC
16 AI capabilities and performance metrics, we should
17 also make certain that this new capability adheres to
18 constitutional standards and privacy rights. I
19 believe maintaining those standards will continue to
20 be a key element of the Intelligence Committee's
21 oversight role.

22 The Director of National Intelligence with

1 input from relevant departments and agencies, bears
2 the responsibility for making sure that the IC tests
3 and safeguards AI systems before deploying them. I
4 should also point out that our nation will face AI-
5 generated threats that not only include direct
6 military threats in all five war fighting domains of
7 air, land, sea, space, and cyberspace, but also
8 include threats to our larger society. Perhaps most
9 importantly, that includes threats to our critical
10 infrastructure.

11 These threats come from nation states,
12 terrorists, and criminal organizations. AI will be
13 used to generate photo, audio, video, and other
14 forgeries of elected officials or other public figures
15 making incendiary comments or behaving
16 inappropriately, the so-called deepfakes. Doing so
17 could potentially erode public trust, negatively
18 affect public discourse, and even potentially sway an
19 election. Such AI-generated products could also be
20 used to embarrass or blackmail elected officials or
21 individuals with access to classified information.

22 A major concern I share with some of my

1 colleagues, particularly senators Young, Schumer,
2 Heinrich, and they're all a group with whom I've
3 worked extensively over the last year on AI policy, is
4 the use of AI to generate novel biological threats.

5 Finally, offensive military use of AI is
6 accelerating the pace of combat. This reduces
7 decision-making timelines for the defender and reduces
8 the opportunity to deter, or if deterrence fails, to
9 defeat an attack. As one of the few members of the
10 Senate who sits on both the Intelligence and Armed
11 Services committees, addressing these AI-generated
12 threats will continue to be a priority. Thank you
13 again for the opportunity to talk to you today. I
14 look forward to a continued dialogue with you in the
15 challenging days ahead for our national security in
16 the age of AI. Thank you.

17 MS. FRANKLIN: Okay. Apologies to our
18 audience for the technical difficulties, but thank you
19 to our IT team for making that work.

20 So, now, hopefully all of our panelists for
21 the first panel will join us, turn their cameras on
22 and welcome to you. Thank you for joining us. So,

1 for our first panel, we will hear from in alphabetical
2 order, I believe, first Alondra Nelson, who is a
3 former acting director of the White House Office of
4 Science and Technology Policy or OSTP. Then Dean --
5 sorry, I'm going to mess up your name, Souleles, yes.
6 former chief technology advisor for the Office of the
7 Director of National Intelligence, then Elham Tabassi,
8 senior scientist at the National Institute of
9 Standards and Technology, or NIST, and then William
10 Usher, senior director for intelligence at the Special
11 Competitive Studies Project.

12 And so for each panelist will make, in order
13 that I just went through, opening remarks up to 5
14 minutes and then we will turn to questioning from the
15 board members. So, Alondra Nelson first. Thank you.

16 MS. NELSON: Good morning. Thank you, Chair
17 Franklin, and members of the Privacy and Civil
18 Liberties Oversight Board. Thank you for convening
19 this critical public discussion on issues associated
20 with the use of AI in the national security context.
21 And I'm honored to be with this distinguished panel.

22 So, I'm a social science scholar and

1 researcher and policy adviser who spent 26 months
2 serving in the leadership of the White House Office of
3 Science and Technology Policy in the Biden-Harris
4 administration. During my OSTP tenure, we stood up
5 the National AI Initiative Office to coordinate AI
6 policy across the whole of government. The National
7 Science and Technology Council that OSTP administers
8 on behalf of the President issued an updated list of
9 critical and emerging technologies, the subset of
10 advanced technologies that are potentially significant
11 to U.S. national security. This list included not
12 only many forms of artificial intelligence, but a
13 number of other technologies that we often consider
14 advanced in part because of their use of systems of
15 data collection, analysis and dissemination that
16 include forms of automation in whole -- in part or
17 whole.

18 At OSTP and my time there, we also launched
19 the National AI Research Resource Task Force, the
20 recommendations of which led to a pilot program to
21 democratize access to the data and compute required
22 for responsible AI development. And we developed, as

1 Chair Franklin mentioned, the blueprint for an AI Bill
2 of Rights, a cornerstone of Biden-Harris AI policy
3 that distills best principles and practices for
4 guiding the safe and responsible design, development
5 and deployment of AI technologies.

6 In my past and current research, I also
7 studied the social implications of science and
8 technology -- of science and technology and related
9 policy and research analysis issues. Across this
10 work, I've come to appreciate that particular
11 challenges that advanced AI presents to both national
12 security, including counterterrorism especially to the
13 -- especially acute regarding the preservation of our
14 principles, norms, and practices we need to protect
15 rights and liberties.

16 AI technologies, both so called Predictive AI
17 and more recent generative AI, have expansive
18 potential use in the national security context and do
19 a lot of work to keep us safe, including intelligence
20 data processing and research, strategic decision
21 making with humans on the loop or in the loop as the
22 case may be, transportation logistics, cybersecurity,

1 there's a growing use of drones, which we should
2 probably discuss, targeting and simulation.

3 One of the examples of use for national
4 defense or planetary defense, moreover, that I often
5 like to talk about is in the space of outer space and
6 international and space policy. You might be familiar
7 with the double asteroid redirection test or the DART
8 mission, which is part of U.S. national and planetary
9 defense. It was designed and carried out to protect
10 Earth from collision with an asteroid or another
11 entity by moving an object out of its orbit and out of
12 therefore a dangerous trajectory. NASA succeeded in
13 this mission for the first time in late 2022. And
14 this was made possible by years of AI-enabled
15 calculation and autonomous simulation, more
16 particularly the Small-body Maneuvering Autonomous
17 Real Time Navigation algorithms or SMART Nav
18 algorithms that allow scientists to predict the path
19 of an asteroid, and then to plan the navigation of a
20 spacecraft to collide with it, and place it on a non-
21 harmful path and also not cause harm to the
22 spacecraft.

1 Crucially important for national and
2 planetary defense, therefore, are -- is something like
3 the DART mission and also is critically important
4 science for the volume of orbital debris, the
5 satellite launches that grow every day, and the kind
6 of geopolitics of space that's happening that poses
7 new national security risks.

8 But I think our discussion today is no doubt
9 about the implications of AI in the national security
10 context prompted by the developments in advanced AI
11 since November of 2022 when ChatGPT was released to
12 the world and the emergence of these kinds of
13 foundation models and what they mean for, as Senator
14 Round suggested, the generation of text, of sound, and
15 image that have been described as general purpose.

16 General purpose, that phrase lies -- herein
17 lies the challenge that AI poses, both the opportunity
18 and the challenge that AI poses for national security.
19 For this new suite of technologies threaten to thicken
20 the so called fog of war, that disorientation and
21 uncertainty of situational awareness in the military
22 theater, they threaten to thicken the fog of war to

1 brattle social effects across both civilian and
2 military domains.

3 So, we might call this potential, the fog of
4 advanced AI, right, and it has a few important facets
5 for our discussion. One, that we are increasingly
6 with advanced AI using inscrutable commercial AI
7 software that can be transformed into many forms that
8 are not fully known. Some of them are quite banal,
9 and some of them might be dangerous, but we don't
10 know.

11 Second and related. The black box that is
12 often necessary for military and IC secrecy with these
13 new inscrutable technologies is compounded and further
14 obscured by an accuracy by biases in the technology
15 and the training data, and by the fundamental weakness
16 of inscrutable technology like generative AI that for
17 many use cases works pretty well a lot of the time,
18 but doesn't work entirely well all of the time.

19 The implications for one and two for the
20 commercial software that can be used for both
21 dangerous and banal uses, that compounds the black box
22 of sometimes necessary military secrecy, means that

1 layered on to defense secrecy is this layer of black
2 box technology that holds significant implications for
3 national security effectiveness and also for public
4 accountability.

5 The traditional notions of dual use
6 technology are technologies that are intended for one
7 purpose and that can have been discovered often to
8 have an application for another use, one purpose being
9 civilian, the other military.

10 A classic case emerging from chemical and
11 biological research has been the development of, you
12 know, bio weapons beginning in the early 20th century.
13 And more recently, we had the development of massive
14 explosive capabilities from the use of ammonium
15 nitrate fertilizer and other chemicals combined that
16 were widely available to carry out the Oklahoma City
17 bombing.

18 This act of domestic terrorism is a perfect
19 analogy for advanced AI and that many civilian and
20 military applications can be made inherently out of
21 the work -- out of generative AI. These can be both
22 intended and unintended use cases.

1 For example, we might take the case of facial
2 recognition technology. We know, for example, from
3 reporting, as Chair Franklin mentioned, this is all
4 widely known information that Clearview AI's facial
5 recognition technology is being used in the Russia-
6 Ukraine war, being used by Ukraine to identify
7 deceased Russian soldiers. Clearview's AI systems are
8 known to be built from scraping websites of civilian
9 data, creating potential rights violations in a
10 civilian context importing these into the theater of
11 war.

12 Without public accountability, and there's --
13 these technologies are often -- also used for public
14 security. So, this is not just one technology
15 intended to use in one domain and used in another,
16 what we face today is the circulation of these
17 technologies back and forth across civilian and
18 military domains simultaneously in ways that create
19 new challenges for oversight boards like this one for
20 policymakers who work both on the civilian and
21 military sides and that raise tensions for democratic
22 societies.

1 Facial recognition technology used
2 domestically by police, including DataWorks Plus in
3 Detroit has yielded numerous cases of
4 misidentification that I bet have had high costs for
5 people's lives, including for Robert Williams, an
6 African American man arrested in front of his family
7 for burglary he wasn't involved with.

8 MS. FRANKLIN: Thank you, if you could just
9 wrap up your opening so we can move on to the other
10 panelists and hopefully have more time for questions.
11 Thank you.

12 MS. NELSON: Sure. Yeah, yeah, okay. So, to
13 date the government has -- what is clear is that the
14 US will need to develop new standards of practice and
15 engagement that do not adhere to the technology not to
16 AI but to the mission and values of the U.S. And this
17 is because these technology, commercial technologies
18 will have to be -- decisions about them will have to
19 be shared not only across the IC, but across the
20 Department of Commerce, FTC and other executive
21 agencies. Public accountability has always been hard
22 to accomplish regarding military uses of technology.

1 But this becomes more urgent in the context of general
2 purpose dual use technologies.

3 With the introduction of advanced AI, we can
4 no longer effectively or neatly separate civilian laws
5 and regulations from military ones. War is often the
6 best -- worst way to preserve a way of life and to use
7 AI in a way that diminishes our basic values is not
8 mission-aligned. Allied countries can work together
9 to minimize abuse by reducing the circulation and
10 dissemination of commercial AI technologies with
11 export controls and sanctions.

12 But fundamentally an unregulated U.S.
13 commercial AI technology industry with dual use
14 general purpose technology increases national security
15 risks. Fundamental regulation is needed. I know this
16 is not the mandate or domain of authority for the
17 board. However, the board can use its sphere of
18 influence to see where the various responsible use of
19 AI exist.

20 MS. FRANKLIN: Thank you. Thank you. I'm so
21 sorry to interrupt you. But I do want to make sure we
22 have time for everybody's opening, and then for the

1 questioning with the board. Thank you so much.

2 Okay, so we will next hear opening remarks
3 from Dean Souleles.

4 MR. SOULELES: Perfect. Thank you so much.
5 Thanks for convening this session. It's a very
6 important session. My role in government and as a
7 career technologist was often at the intersection of
8 technology and management or mission, and to translate
9 for the technologists what the actual mission is, and
10 to translate for the mission what the technology is
11 and what it is and what its limitations are. So, I
12 kind of sat at that intersection in my time at the
13 Office of Director of National Intelligence. And I
14 want to talk a little bit about that, from that
15 perspective.

16 AI is clearly very important to the
17 counterterrorism mission but, as always, I'd like to
18 start with defining our terms. So, when I speak of
19 the counterterrorism mission, I'm speaking very simply
20 of our mission to collect, analyze, and share
21 actionable intelligence related to terrorism and to
22 detect and disrupt those threats. So, within that

1 context, we need to look at what AI is and what it
2 isn't.

3 And, in addition, at the Office of Director
4 of National Intelligence, the National
5 Counterterrorism Center, is responsible for
6 maintaining the authoritative database of known and
7 suspected terrorists. So, we got a big database of
8 people. That's a identification issue. So, that's
9 the ICCT mission. But what do we mean by AI? And if
10 you have a conversation about AI and civil liberties,
11 you better know what you're talking about. And that's
12 not so easy to answer.

13 In the current environment, you could be
14 excused for thinking that AI is synonymous with large
15 language models and chatbots. If you haven't been
16 deeply involved in technology, it appears that this
17 technology came out of nowhere 2 years ago. Well, it
18 didn't.

19 And it's now seemingly everywhere, it's
20 pervasive. But this is the latest in a long, long
21 line of machine intelligence tools that have become
22 increasingly and more useful over the last decade.

1 And by the way, the DOD and the U.S. Intelligence
2 Community have been using these tools for years, many,
3 many years. This is just the latest in a set of
4 technologies.

5 In 2012, Yann LeCun and Geoffrey Hinton
6 demonstrated neural network based supervised machine
7 learning was better than or equivalent to human in
8 many cases. And that triggered this wave of
9 technology that turned into the kinds of technologies
10 that we are seeing today. I broadly classify AI tools
11 into a bunch of buckets, there's a bunch of different
12 taxonomies. But a useful one is to think about
13 supervised machine learning. This is where we take
14 large amounts of ground truth data, which is called
15 training data, usually provided and curated by human
16 experts into a machine classification system. That's
17 how image recognition and facial recognition works.

18 And as we've heard, if you put the wrong data
19 in, you're going to get the wrong conclusions out,
20 it's going to have bias. Then there's unsupervised
21 learning, which takes massive amounts of data and
22 seeks to find patterns or connections in the data in a

1 way to make it more useful.

2 Again, it's only going to produce relevant
3 insights based on the data that has been fed. And we
4 make a bias decision every time we choose what to
5 include, or what not what not to include in those
6 systems. And what questions to ask of those systems.

7 Another kind of AI is reinforcement learning.
8 And this is a set of AI technologies where the system
9 learns how to behave in ways that increase reward,
10 they call it -- mathematicians call it a reward
11 function, by interacting with the environment. In
12 other words, the AI gets it right, you increase the
13 reward, which is a numerical number, if it gets it
14 wrong, you decrease it. And you run these things many
15 tens of thousands or millions of times and that's how
16 you get a computer that can beat the best players at
17 Chess and Go with this idea of reinforcement learning.
18 That's not as intended, that's not dependent on data,
19 it's really dependent on a set of rules. But you can
20 bias that system however you like, by choosing the
21 rules in which you wish to train it.

22 Deep learning underlies all that. And it's a

1 set of technologies that work across all the areas I
2 just talked about. It uses large quantities of data
3 to figure out how to do complex things, searching
4 through combinations of ways that best describe the
5 data. So, that's kind of the context of this. And
6 all of those things, all the things that we think of
7 as AI are one of those sorts of things. And they're
8 basically computer decision making, computer search,
9 advanced decision making advanced analytics, there's
10 all kinds of ways you can describe it. But at the end
11 of the day, they are mathematical models that help us
12 make conclusions about data.

13 We may have done ourselves a disservice by
14 personifying things like ChatGPT and having it speak
15 in human terms. These are not humans, they are not
16 brains, they do not think, they do not have hopes and
17 dreams, you turn them off, and they don't -- they go
18 to sleep and they don't come back. We need to be
19 aware of what it is we are talking about. So, when we
20 set up the AI strategy for the IC, I felt it was very
21 important that we address things like what are the
22 risks, and we know that AI can learn the wrong thing,

1 if it's given the wrong data, we know it can do the
2 wrong thing. And worse, it can do it with confidence.
3 It will always give you an answer.

4 And you need to be aware as a human analyst,
5 that the fact that it gives you an answer isn't the
6 fact that it is correct. And it can even reveal the
7 wrong thing. So, in the context of national security,
8 if our models, our classified black box models leak
9 out into the world, we know that as analysts we can
10 analyze those models and learn what training data that
11 we were training them on. So, these are risks. Now
12 there are a huge number of other things that we can
13 talk about, and Alondra mentioned many of those. I
14 won't repeat them. But I'm happy to happy to take any
15 of your questions in the question round.

16 MS. FRANKLIN: Thank you. Okay, so we will
17 next hear from Elham Tabassi.

18 MS. TABASSI: Good morning, everyone.
19 Grateful for the invitation, chair Franklin, member of
20 the Boards. I'm delighted to be here among
21 distinguished speakers. Always difficult to follow
22 Dr. Nelson and also Dean, but I try to do my best.

1 Again, thanks for the opportunity to come and talk
2 about some of the things that we have done in the
3 space of AI and AI risk management.

4 For some of the audience that may not know
5 NIST, National Institute of Standards and Technology,
6 we are a measurement science agency. NIST is a
7 nonregulatory agency under Department of Commerce with
8 a unique mission to advance U.S. innovation. We have
9 a very broad portfolio of research at NIST, but more
10 importantly, a long tradition of cultivating trust in
11 technology. And we do that by advancing measurement
12 science and standards, measurement science and
13 standards that makes technology more reliable, secure,
14 private, fair, in other words, more trustworthy. And
15 that's exactly what we have been doing in space of AI.

16 NIST was established in 1901 to fix the
17 standards of weights and measures. Our predecessors
18 created advanced standards to measure basic things
19 such as length, mass, standards needed for
20 electricity, light, everything that was essential for
21 the technological innovations and competitiveness at
22 the turn of the 20th century. And in a way, we are

1 following the same course working with and engaging
2 the whole community in figuring out proper standards
3 and measurement science for advanced technologies of
4 our time, which I think everybody agrees artificial
5 intelligence is in that category.

6 In terms of what we have been doing in this
7 space, a little over a year ago, something like a
8 year-and-a-half ago, we released a NIST AI Risk
9 Management Framework or AI RMF. Directed by
10 congressional mandate, AI RMF is a voluntary framework
11 for managing the risk of AI in a flexible, structured
12 and measurable way. The measurable attribute is
13 particularly important for us, coming from a
14 measurement science agency, because if we cannot
15 measure it, we cannot improve it.

16 So, if you want to really improve the
17 trustworthiness and responsible use of AI, we need to
18 be able to have measure -- to know what to measure and
19 how to measure. AI RMF was developed in close
20 collaboration with AI community, we engage diverse
21 groups of different background expertise and
22 perspectives from the community that developed the

1 technology to the community that study the impact of
2 the technology to running listening sessions with a
3 community that are impacted by the technology. The
4 framework is intended to be voluntary, rights-
5 preserving, nonsector specific and use-case agnostic,
6 providing flexibility to organizations of all sizes in
7 all sectors and throughout the society, to implement
8 the approaches in the framework. So, by design, it
9 can be used for all of those different application
10 that Dr. Nelson mentioned, and Dean also alluded to
11 them.

12 Continuing that work in March of 2023, we
13 released AI Resource Center as sort of a one-stop-shop
14 of knowledge, data, tools for AI risk management. It
15 houses AI RMF playbook that provide more sort of
16 actionable suggestion on how to implement and
17 operationalize AI RMF. It's cool, it's interactive,
18 searchable, filterable. And we consider that as a
19 work in progress as we're adding additional
20 capabilities. For example, things such as standard
21 hub or repository of metrics are more.

22 In June of 2024, again in response to the

1 release of the generative AI languages, we put
2 together a generative AI public working group where
3 more than 2,000 volunteers helped us to sort of study
4 understand the risks that are unique to generative AI
5 or exacerbated by generative AI. Our latest
6 assignment, the executive order on safe, secure, and
7 trustworthy AI builds up on all of those foundational
8 work that we have been doing. The executive order
9 specifically directed NIST to develop evaluations,
10 red-teaming, safety and cybersecurity guidelines,
11 facilitate development of consensus-based standards,
12 and provide testing environment for evaluation of AI
13 systems, including for dual use foundation models.

14 All of these guidelines and infrastructures
15 will be voluntary resources for the use by AI
16 community for advancing safe, secure, and trustworthy
17 AI. I think it has been mentioned several times, I --
18 it's -- everybody knows that AI is the, one of the
19 most transformative technologies of our time, one with
20 tremendous opportunities to improve our lives, but
21 also comes with its negative consequences and harms.
22 That's why safeguards becomes really important.

1 When it comes to AI, there is a lot less we
2 know that we should, and I think all of these
3 conversations and what we can do is that we should try
4 to change that. There is a lot that we can do. I'm
5 just going to talk about five things that I jotted
6 down last night.

7 So, first, we heard it in different ways that
8 our understanding of limits and capabilities of this
9 powerful technology is limited, so we must engage in
10 efforts, technical and scientific efforts to advance
11 our scientific understanding of how these models work
12 and behave.

13 We heard this, this morning. But I also want
14 to emphasize that we also must address AI's impact on
15 people and society and planet through technical,
16 social, and sociotechnical lenses. We should also
17 advance research on identifying, measuring, managing,
18 and mitigating risks, including safety, security,
19 privacy, fairness, reliability, interpretability. One
20 of the things AI RMF does is try to provide some sort
21 of a taxonomy of the risks for AI systems to help with
22 this structured, measurable approach to risk

1 management.

2 You should also, and I think this is really
3 important, actively seek and incorporate insights from
4 a diverse range of experts representing diverse set of
5 backgrounds and perspectives, particularly the group
6 that the technology is going to impact them. And
7 data, technology does not know borders, so it's
8 important to cultivate and strengthen international
9 collaboration, cooperations on AI issues, but
10 particularly on standards. Bottom line is that we
11 want technologies that work accurately, reliably,
12 technologies that's easy to do the right thing,
13 difficult to do the wrong thing, and easy to recover
14 if and when something goes off. And --

15 MS. FRANKLIN: If you could please wrap up
16 your opening, so we do have time to get to the
17 questions and answers.

18 MS. TABASSI: I think that's a good stop.
19 Good place to stop. Thank you.

20 MS. FRANKLIN: Thank you so much.

21 Okay. And our final opening will come from
22 William Usher.

1 MR. USHER: Good morning, distinguished
2 members of the board and for those listening, and I'll
3 keep my remarks brief so we can get to the question-
4 and-answer period.

5 Again, my name is William Usher. I'm the
6 Senior Director for Intelligence here at the Special
7 Competitive Studies Project. Our mission at SCSP is
8 to make recommendations that strengthen America's
9 long-term competitiveness on emerging technologies as
10 they reshape geopolitics and society over the coming
11 decade.

12 Prior to joining SCSP last year, I spent 32
13 years as an all-source analyst and a senior executive
14 with the Central Intelligence Agency. And I'm honored
15 to speak with you today about the role that AI plays
16 in the national security arena, specifically with
17 regard to the intelligence community's mission.

18 As Ms. Tabassi just said, artificial
19 intelligence stands out as a transformative force that
20 will profoundly impact national security and global
21 competition. President Biden's executive order last
22 October mandated that the U.S. government departments

1 and agencies take care of when developing and
2 deploying AI systems. But it also called on America
3 to "seize the promise of this powerful new
4 technology."

5 Being a leader in technology innovation is
6 important today, but it will be vital to our nation's
7 future economic vibrancy and to the continued
8 resiliency of our democratic way of life in the
9 future.

10 As we debate the future of AI, foreign
11 competitors, principally the People's Republic of
12 China, are laser-focused on taking advantage of AI for
13 economic advantage, and to challenge U.S. leadership
14 and the rules-based order. Beijing has openly
15 declared its aspiration to become a leading S&T power
16 that is able to set the pace of future scientific
17 advancements and dictate global norms.

18 Now, our intelligence community has long eyed
19 AI's potential, and they have been researching the
20 potential uses of early forms of AI, machine learning,
21 deep learning and natural language processing for
22 years and have already launched limited uses of

1 generative AI tools.

2 Gen AI tools have the potential to greatly
3 expand the scale and the efficiency with which our
4 intelligence services can derive national security
5 relevant insights from the growing body of digital
6 information produced around the globe.

7 U.S. intelligence services, for example, will
8 be able to leverage AI's pattern recognition
9 capabilities to identify and alert human analysts to
10 threats such as potential terrorist attacks, or
11 significant military movements. This capability will
12 make critical warnings more timely, actionable, and
13 relevant, allowing for more effective responses to
14 emerging threats and hidden strategic opportunities.

15 While the potential is great, AI also poses
16 significant new challenges for our national security
17 enterprise. For one thing, a host of foreign
18 countries, including several U.S. adversaries, are
19 already investing heavily in AI for their own national
20 security purposes.

21 China, for instance, is expected to more than
22 double its investment in AI to nearly \$27 billion by

1 next year and \$38 billion by 2027. Moreover, there is
2 a great deal of -- while there is a great deal of
3 attention being paid today to the creators of large
4 expensive-to-train foundation models, the presence of
5 several capable so-called open-source models, I'm
6 thinking here of Meta's Llama or Mistral 7B model
7 means that the speed at which this technology will
8 become generally available will be very rapid. Unlike
9 the advent of the atomic age, you will not need to be
10 a well-resourced nation state to be able to benefit
11 from AI technologies.

12 Therefore, our intelligence services must
13 devote additional resources and effort to ascertain
14 what foreign competitors and non-state actors are
15 doing to develop their own indigenous AI systems, and
16 how they intend to employ them against us and our
17 allies. We've already seen evidence of AI being used
18 to create believable misinformation, lifelike videos
19 and audio files that appear authentic, that are being
20 used to push false narratives. But these same AI
21 tools can be used to uncover sensitive U.S. military
22 and intelligence operations, plan more sophisticated

1 cyber attacks, and develop novel bio weapons.

2 It is this Board's mandate to provide
3 oversight of the Federal Government's implementation
4 of the AI executive order and this certainly poses
5 some important new questions that need to be
6 addressed. The forthcoming White House National
7 Security Memorandum likely will provide the initial
8 framing of how the government thinks these questions
9 should be properly answered when it is released later
10 this month.

11 To my mind, those questions fall into one of
12 two broad categories. The first category is what are
13 the parameters that will guide whether the IC can make
14 use of any particular model. If leading-edge large
15 language models are basically trained off the
16 internet, which is composed mostly of U.S.-derived
17 information, how does that affect IC agency's use of
18 such models?

19 Specifically, how can agencies utilize AI and
20 remain compliant with Intelligence Community Directive
21 107 concerning privacy protections? My personal view
22 is this can be done but right now I think different

1 agencies are interpreting the rules differently. The
2 second category is what will be non-acceptable uses of
3 generative AI outputs for the U.S. intelligence
4 community.

5 As we try to figure that out what are the
6 examples of non-acceptable uses, I expect we will go
7 through a lengthy trial and error process and formed
8 mostly by "I'll know it when I see it" type wisdom.
9 Some restricted areas will be obvious, such as relying
10 solely on AI systems to target suspected terrorists
11 for kinetic strikes, but other potential restrictions
12 will be less obvious.

13 For example, imagine a scenario in which a
14 U.S. intelligence service proposes to request that
15 another government detain a foreign national
16 transiting their country, which the intelligence
17 service assesses is engaged in a terrorist plot based
18 purely on the recommendation of an LLM AI model. What
19 are the expectations for human review of that
20 recommendation?

21 Or more challenging, what if the AI detects
22 what it assesses to be an imminent cyberattack that

1 could occur any second? The AI tells you it knows
2 exactly which U.S. computer systems to lock down to
3 thwart the attack. There is no time to gather
4 policymakers for a meeting in order to head off the
5 attack. Is the AI pre-authorized to mount a defense?

6 We'll see what the National Security Memo
7 says and whether a clarify thing -- clarifies things
8 or not. But I suspect we are embarked on a long
9 journey to determine whether and more importantly, how
10 the IC uses AI to its advantage. I recognize the
11 risks. But I would encourage the President, Congress,
12 and this Board, not to prematurely tie the IC's hands
13 because our adversaries certainly are making use of AI
14 and we need to stay ahead. And with that, I'll close
15 my remarks and take any questions.

16 MS. FRANKLIN: Thank you so much. Thank you
17 to all of the panelists. So, we are going to try to
18 cycle through twice, hopefully, with all board members
19 having a chance to ask questions. And I am kicking
20 off this round.

21 And I want to start with Alondra Nelson,
22 please. So, you discussed the White House's Blueprint

1 for an AI Bill of Rights that I believe you lead for
2 OSTP. And in the context of data privacy, that
3 Blueprint lays out the need to limit data collection,
4 to follow privacy by design principles and to
5 incorporate robust safeguards, excuse me, robust
6 oversight for automated systems.

7 And in particular, I noted that the Blueprint
8 describes the need for heightened oversight of
9 surveillance systems, including an assessment of
10 potential harms, both before deployment and in an
11 ongoing manner and to test for harm such as
12 algorithmic discrimination. So, I'm wondering, can
13 you provide us with any further thoughts or more
14 detailed guidance on how in your view government
15 agencies should conduct these pre-deployment
16 assessments and what kind of research you've seen and
17 what promise that holds in that space?

18 MS. NELSON: Thank you for that question,
19 Chair Franklin. So, in the process of developing the
20 Blueprint for an AI Bill of Rights, effectively what
21 we did was distill best practices from industry, from
22 academia and from colleagues working in government,

1 about these technologies. And so, you know, what we
2 distill there is what we've learned from what people
3 think is possible, or what they've been already using.

4 In the space of government and in particular,
5 you know, obviously the pre-deployment assessment will
6 happen in the space of acquisition and procurement.
7 And there's quite a lot in the President's executive
8 order, I think that, you know, that asks agencies to
9 think about that piece of their work, as well as, we
10 haven't mentioned yet OMB's memo on the trustworthy
11 and safe use of AI, which, you know, suggests, as we
12 would want to, you know, that government should be in
13 the business, should be leading by example, and in the
14 business of using rights-preserving technologies and
15 technologies that if they have impact on people's
16 safety, that we're thinking, you know, about how to do
17 that.

18 I would say that the Blueprint for an AI Bill
19 of Rights, depending on whether or not you read the
20 PDF, or the website, has an IC carve out. And so, you
21 know, the sort of the principles are, you know, of the
22 -- Blueprint for an AI Bill of Rights and the

1 practices, as you suggest, the, you know, pre-
2 deployment, the sort of various assessment tools that
3 people might use, are not intended to apply to that
4 space.

5 What I was trying to suggest in my remarks,
6 as I was closing, and I apologize for running over was
7 that, you know, as SCSP was describing in a world in
8 which, you know, new threats can come from, you know,
9 actors using these commercially -- widely commercially
10 available technologies, that one of the ways that we
11 can exercise, national security kind of prudence and
12 oversight is actually to have oversight of commercial
13 technologies in the civilian sphere.

14 And, you know, that's where I think this
15 Board's ability to exercise oversight over the
16 fulfillment of the executive order and other kinds of
17 executive agency, you know, sort of mandates and
18 levers is tremendously important.

19 MS. FRANKLIN: Thank you. Okay. So, I'd
20 like to ask a question of all the panelists to
21 hopefully quickly touch on this for us. I know, it's
22 a big question though, which is, you know, as I noted

1 in kicking off this forum, we are working to scope and
2 define our oversight of governments to use of AI for
3 counterterrorism purposes. And I would appreciate
4 any, you know, sort of concise thoughts you may have
5 for us on how we should carve out an appropriate
6 slice.

7 And where I'm going with that is, you know,
8 of course, it would be completely unworkable for us to
9 say, okay, right now we're going to start examining
10 all the government uses of AI for counterterrorism
11 purposes, or even to say all surveillance programs, or
12 even to say all data analysis. So, we want to be
13 strategic, and to think particularly about uses of AI
14 that are more likely to present risks to privacy and
15 civil liberties.

16 So, maybe I'll walk through the order,
17 starting with Dean Souleles, if just any quick
18 thoughts that you want to share with us on how you
19 would advise us to carve out where we go next.

20 MR. SOULELES: Yeah. I think it's unworkable
21 to dive down to any etches. I think instead what I
22 would focus on is ensuring that the intelligence

1 agencies that carry out the counterterrorism measures
2 have appropriate policies and oversight in place to
3 manage their AI systems. So, we developed an AI
4 maturity model for the Intelligence Community while I
5 was still in the seat that basically told the IC how
6 to evaluate its readiness to deploy AI systems.

7 And that includes things like, do you have a
8 data -- chief data officer who's responsible for
9 Privacy and Civil Liberties Oversight? Do you have
10 policies and procedures in place that are such that
11 they force you to analyze the data that you're
12 collecting and ask all the questions that we've raised
13 here today?

14 Do you understand where your algorithms come
15 from? Do you understand the models? So, these are
16 basic things that they must do. And I think the
17 Oversight's Board role is to make sure that they are
18 doing the things that they say they're doing, kind of
19 all right, you've got this set of standards
20 demonstrate to us that you are actually doing the
21 things that you are saying you are doing because I
22 don't think you're going to have the ability to get

1 down any of the issues, but I think if you look at it
2 for the macro level to make sure they have the
3 policies in place, that the policies are consistent
4 with the President's guidance and so on, that you'll
5 have a chance of actually doing what you're asking.

6 MS. FRANKLIN: Thank you. Okay. Quick
7 thoughts from Elham Tabassi.

8 MS. TABASSI: Seconding and echoing
9 everything that was mentioned. And, you know, we are
10 a big fan of the risk-based approach. And considering
11 the contexts of use, there is no one-size-fits-all.
12 So, having the framework, having the question that
13 needs to be asked, but having the flexibility to sort
14 of set the threshold of how private is private, how
15 bias is bias, set them based on the risk of that
16 particular context.

17 MS. FRANKLIN: And quick thoughts from
18 William Usher?

19 MR. USHER: Sure, Dean's got it exactly
20 right. Right now the Intelligence Community for its
21 human employees has clear and firm standards for their
22 use of data and information for various purposes. And

1 your Board provides oversight of that activity. And I
2 think that should be the same standards really, that
3 are used for monitoring how they use AI.

4 I would hope that that does not extend to
5 preventing the IC from taking in large language models
6 for experimentation and examination. But certainly
7 when it comes to utilizing the outputs from LLMs,
8 applied to any classified data holdings, the same
9 standards that are in place now for protecting privacy
10 and civil liberties should apply.

11 MS. FRANKLIN: Thanks. Okay. So, I'm seeing
12 that my time is up. So, I'm going to turn this over
13 to Ed Felten.

14 MR. FELTEN: Thank you. And thanks to all of
15 the panelists for your remarks and your willingness to
16 entertain our questions. I'd like to ask all the
17 panelists about something that has been mentioned a
18 couple times already. And that is about the use of --
19 potential use of foundation models in intelligence and
20 law enforcement. These, you know, as you know, are
21 the largest and most sophisticated of AI models, and
22 they're behind products like ChatGPT, they seem to

1 have unique capabilities that may be valuable for
2 national security missions.

3 But of course, training them requires
4 enormous investments and huge volumes of data. And
5 this is typically done by commercial parties. But
6 government agencies of course have detailed and strict
7 limits on the use of information, and for good reasons
8 relating to privacy and civil liberties. So, my
9 question is really whether and how intelligence
10 agencies might approach the use of foundation models,
11 in a way that's consistent with privacy and civil
12 liberties. Should agencies work with commercially
13 trained models? Should they seek to build their own
14 models? Is there some other approach? Or is this
15 just a bridge too far from a privacy and civil
16 liberties standpoint?

17 Let me start with Mr. Usher, and then go
18 backward in order of the initial -- reverse order of
19 the initial statements.

20 MR. USHER: Mr. Felten, that's a terrific
21 question. What we have recommended is that, yes, the
22 IC does make use of commercially available models,

1 because, as you noted, the cost for developing
2 independently is going to be quite steep and I would
3 argue probably prohibitive. I have had the pleasure
4 of seeing one instance of the IC's use of a large
5 language model as applies to unclassified data.

6 And in thinking about how it could be
7 deployed, how they could be deployed against
8 classified holdings, I think what we would probably
9 expect to see as the capabilities curve kind of goes
10 up with these models in the years ahead, the IC will
11 kind of have to pick a version and work with it, train
12 it to its standards for protection of civil liberties,
13 also tradecraft standards, accuracy, et cetera. And,
14 except that it will be a little bit behind the leading
15 edge of what some of these foundation model developers
16 are doing with their systems, but be more confident in
17 their reliability, transparency, explainability, et
18 cetera because the standards that the IC must meet for
19 telling the truth and protecting civil liberties is
20 and should be high.

21 MR. FELTEN: Thanks. Let me turn to a Dr.
22 Tabassi.

1 MS. TABASSI: I completely agree, nothing
2 more to add here, just saying that, yes, you can get
3 the model and then try to add extra safeguards to make
4 it up to the thresholds of the IC.

5 MR. FELTEN: Right. Mr. Souleles?

6 MR. SOULELES: Yeah, I would say that it is a
7 great question, but it is the same question that we
8 ask with all deployed technologies and all data
9 analytics. It's a little more complex because, as we
10 know, these foundation models are trained on huge
11 volumes of data. And in the Intelligence Community,
12 that means they're trained on data that involves U.S.
13 persons.

14 I spent a good deal of time having this
15 conversation with our attorneys and others in terms of
16 how do we use the tools without violating the basic
17 directives that were not allowed to perform
18 intelligence on U.S. persons. And I would say that we
19 need a policy that talks about that specifically and
20 makes it clear what we can do and what we cannot do.
21 I think we should be able to use foundational models.

22 As SCSP said, we should retrain them to our

1 own standards, but we should not eliminate the use of
2 them, because they have been trained on U.S. person's
3 data, because that would cut us off from a wide swath
4 of technologies that we know our adversaries are
5 using. And this is really similar to the conversation
6 a few years back with NSA and telecommunications
7 election, right, is, yeah, there's going to be U.S.
8 person's data in there, it's incidental to the
9 intelligence problem that we're asking. And we
10 develop policies and procedures for using it. We
11 didn't throw it all out completely. So, my approach
12 would be to embrace it, put the appropriate guidelines
13 around it and continue to use it.

14 MR. FELTEN: All right. Dr. Nelson?

15 MS. NELSON: Thank you, Dr. Felten. A few
16 things. I mean, I raised the Clearview AI example
17 because it suggests some of the challenges that we
18 faced here. So, there's clearly American data in
19 there. It's being used by Ukraine and Russia in the
20 theater of war. And there's a lot of complexities, I
21 think that we still need to think through in the
22 national security space with regard to how we as

1 Americans want to operate in using that.

2 Second, I would say is that there's a -- you
3 know, we're seeing a kind of a lot of David and
4 Goliath, if you look at the example of what's
5 happening in Gaza right now. So, having a big model
6 is actually not necessarily going to be the thing that
7 helps you win, if you like, you know, switching back
8 over to Ukraine, when people can take commercially
9 available drones, and, you know, that cost \$1,000 and
10 destroy a multimillion dollar tank using that, right?

11 So, I think that the, you know, smaller
12 language models, the open-source models create a kind
13 of asymmetry that we want to use the foundation
14 models, I think, safely and effectively knowing all of
15 the many, many caveats around them. But I think that
16 a lot of what AI is enabling is this kind of radical
17 asymmetry in the national security space.

18 And then lastly, I would say I would just
19 point people to and commend that DARPA has just
20 started a new program on the mathematical foundations
21 for AI evaluations. And I think one of the first
22 things that the IC needs to do is actually to Elham's

1 point, figure out how these things actually work
2 because just air gapping the data alone or using the
3 enterprise version, I don't think for the threshold
4 you want for national security is actually high
5 enough. And so I think moving forward on
6 understanding the basic science of this is
7 tremendously important as well.

8 MR. FELTEN: Great. Thank you. Thank you to
9 all the panelists for your thoughtful answers. Let me
10 pass the baton to my colleague, Travis LeBlanc.

11 MR. LeBLANC: Thank you, Ed. And also, I
12 want to thank the panelists for joining us this
13 morning for this important forum on artificial
14 intelligence and how we balance it with privacy and
15 civil liberties in the national security context.

16 I'd like to pick up on a conversation that
17 Dr. Nelson was just having around Clearview AI. And I
18 do appreciate the concerns that you elaborated on
19 about the use of Clearview AI by Ukraine. It has been
20 called Ukraine's secret weapon in the war. And the
21 question I sort of have is, do you believe that the
22 United States should refrain from using an application

1 or tool like Clearview AI? And if so, what do you say
2 to those who believe that it puts the country at a
3 disadvantage to defend itself if our adversaries are
4 able to use these tools?

5 MS. NELSON: Thank you for that question, Mr.
6 LeBlanc. We already use it widely. It's used by
7 American police forces all over. So, it's not that's,
8 you know, that's in some ways it's a moot question. I
9 think it gets a little bit more complicated in the
10 international sphere when we're talking about civil
11 liberties and people's rights when it's American data
12 that's being deployed in the theater of war and other
13 spaces in Ukraine and in Russia and how do we want to
14 think about that at a time when, you know, this
15 administration is issuing executive orders that is
16 constraining the flow of data for example. So, that
17 American data should not be allowed to circulate in
18 countries of concern for example.

19 So, to me the Clearview AI issue I raised
20 because it raises a lot of fundamental questions that
21 we don't have answers to and a lot of fundamental
22 tensions. So, Mr. LeBlanc, I don't have any clear

1 answers. But I would say we're already using the
2 technology, gets a lot more complicated when these
3 commercial technologies are also become, you know,
4 military technologies. And then we need to reimagine,
5 I think the regulatory and rights regimes or either
6 double down on them and, you know, we've got to figure
7 that out.

8 MR. LeBLANC: And I guess thank you for that
9 response. I want to follow up to ask, are there any
10 applications of AI or any uses of AI that you believe
11 the U.S. government should not be engaged in right
12 now?

13 MS. NELSON: It depends on the context. I
14 don't think that we should be using real-time facial
15 recognition technology in a civilian context at all.
16 I think that should be a red line.

17 MR. LeBLANC: Okay. Thank you very much.
18 And, you know, we have a lot of professors on this
19 webinar, but we only have one Dean. So, I want to ask
20 Dean one question, which is, is the error rate around
21 generative AI too high right now to be reliable? And
22 relatedly, are there any uses of AI that should be

1 halted?

2 MR. SOULELES: So, the error rate of
3 generative AI is pretty high. If you're talking about
4 large language models and what we are all seeing out
5 on the internet as chatbots, it's very easy still to
6 do a search, do a query to the chatbot and have it
7 return, you know, what they call hallucinations. And
8 the important thing to understand is that despite the
9 marketing of these tools, these are not knowledge
10 systems, they are predictive text systems, they are
11 trained. The idea of a large language model is it is
12 trained on essentially the entire text of the
13 internet. And it's able to produce in a remarkable
14 way an English-readable sentence and paragraph and
15 sentences based on all the text that has already been
16 produced. But it doesn't think in any of the ways
17 that humans think. So, we should be very careful when
18 we use those sorts of things.

19 I would say though that intelligence analysts
20 are already used to working in a probabilistic world.
21 They have to have data that is checked against other
22 data. They never take one source of data as the

1 ground truth. So, I would say I wouldn't prohibit the
2 use of them. But we need to understand how they're
3 being used and we need to not use them as a source of
4 ground truth. Just like I wouldn't use Wikipedia as
5 my ground truth for data. It's a, you know, crowd
6 sourced encyclopedia.

7 So, are there areas where I would say we
8 should not use it? I mean, when you get down to
9 decisions of targeting individuals for lethal action,
10 then we cannot see that to the automation today, and
11 probably not ever, right? Bob Work talks about the
12 need to have AI implement commander's intent in
13 warfare, just as we implement commander's intent when
14 we issue orders to troops. And the same rules of the
15 road should apply.

16 But at some point, the gap between the
17 commander and the execution of that command is broken.
18 And you have to rely on the thing that's executing the
19 command to do its thing to make sure that it's built
20 that way. So, hope that's helpful.

21 MR. LeBLANC: That is indeed helpful. And,
22 you know, your reference to targeting and, you know,

1 prohibitions on targeting individuals for lethal
2 action reminds me for one last question for Dr.
3 Nelson, which is, you did say in your opening
4 statement that we had to talk about drones. And I
5 just want to give you a moment in case you'd like to
6 discuss any of the concerns or other issues that you'd
7 like to cover around drones.

8 MS. NELSON: Thank you for that, Mr. Leblanc.
9 So, I think drones is also another case study for
10 thinking about the challenges we face at the
11 intersection of the national security and civil
12 liberties piece. You know, as I said, these are
13 relatively inexpensive technologies. As we're seeing
14 in Ukraine, they're being kind of refitted with, you
15 know, cameras and other things to be used for -- to be
16 made sort of as semi-autonomous weapons, you know,
17 with that are partly guided.

18 And so, then we have growing capabilities out
19 to swarm drones to have them act as both as agents and
20 collectively. And so, then that runs into questions
21 that we have around, you know, lethal autonomous
22 weapons, right, like conventions that are existing

1 around that, you know, regulations, and the
2 international relations space that exist around that
3 and what we might need to think about and new ways
4 about that.

5 And, you know, Ukraine is an interesting
6 example because it's already been, and I'm sure will
7 continue to be, a really important technology capital.
8 There's great technologists there. And so, part of
9 what we're seeing is about the capabilities of this
10 particular community to be able to take drones and
11 make them into warfare. But I guess the challenge
12 that this board faces, I think, is that back and
13 forth, those back and forth vectors between civilian
14 and military technologies that make these questions
15 open questions, rather than I think clear-cut answers.

16 MR. LeBLANC: Thank you. And do you think
17 that there would be a good use of the Board to look
18 into how DOD uses artificial intelligence in lethal
19 strikes?

20 MS. NELSON: Yes. I mean, I think others,
21 either Dean or SCSP had -- has already mentioned this.
22 I mean, the DOD has been quite a leader. And when I

1 first came to OSTP, as a day one person in the Biden-
2 Harris Administration, I believe DOD was one of the
3 few agencies that had already released a set of
4 principles and guidelines around AI. So, like very
5 forward-leaning here. And the question then becomes,
6 you know, how do we ensure that people are actually
7 doing that, which colleagues have already mentioned?
8 And I think that's a perfect place for this particular
9 board to exercise its oversight.

10 MR. LeBLANC: Thank you. Beth Williams, I'm
11 passing the baton to you.

12 MS. WILLIAMS: Okay. Good morning. Thank
13 you, Travis. And thank you to all of our panelists.
14 Really appreciate you being with us here today for the
15 forum.

16 So, my first question is actually to Mr.
17 Usher. So, one question is how can the intelligence
18 community leverage commercially available and open-
19 source resources and still protect classified
20 information that's used for developing, deploying, and
21 using its own in-house AI systems? One of the things,
22 you know, in the reading was the concern that AI

1 systems that are in use by the IC could be reverse-
2 engineered to divulge classified sources. And that's
3 obviously a big concern, not only for national
4 security, but for the privacy and civil liberty and
5 safety of the sources themselves. So, could you talk
6 a little bit about that?

7 MR. USHER: Absolutely. And this is a
8 terrific topic for exploration because as these
9 systems become more and more capable in the years
10 ahead, I predict that they will be viewed eventually
11 as critical national security assets. Some have made
12 the analogy to the Manhattan Project. They will be so
13 valuable that we'll have to, you know, bury them in a
14 deep vault and protect them with several rings of
15 security.

16 Mechanically, today, the way that that is
17 done is by putting them on secure servers, which have
18 built-in protections. It's how we onboard whatever
19 piece of software or data that we wish to use in the
20 Intelligence Community and keep it protected. And
21 they're pretty well-established security protocols.
22 Access to those systems are limited to people with a

1 security clearance, et cetera, et cetera.

2 You mentioned how adversaries will be viewing
3 them. And I certainly think they will be seen as
4 targets, probably targets by adversaries' own AIs.
5 So, one thing to think about is sort of an AI versus
6 AI intelligence for where their AIs are seeking to
7 gain access to our secure servers to pollute,
8 debilitate or otherwise wreck our AI systems. This is
9 a growing area of research known as adversarial AI.
10 And there are various techniques that one could use to
11 attack another's AI systems.

12 You could mess with the training. You could
13 mess with the data. You could give it instructions in
14 the algorithm to generate false or misleading outputs.
15 And there are -- any number of techniques and the
16 entire AI stack that we would deploy for intelligence
17 purposes will need to be protected. And that will
18 include physical protections, protections for the
19 personnel who have access to it and certainly
20 protections for the algorithm and the data.

21 MS. WILLIAMS: Thank you very much.

22 My next question is for Mr. Souleles. One

1 question that our Board looks at when we're doing
2 oversight projects or looking at systems is, what is
3 the value? So, are there systems -- are these systems
4 producing value that justifies their use? And you in
5 your opening statement, you mentioned that AI facial
6 recognition and other biometrics are becoming
7 increasingly useful for keeping track of known
8 terrorists who are trying to obfuscate their persons.
9 And so, I'm interested in that in the context of
10 biometrics in aviation.

11 How do you look at that system? Right now
12 the system is not comparing faces to any terrorist
13 database. Do you think that that is a concern for the
14 overall usefulness of the system?

15 MR. SOULELES: I do. I mean, I think that we
16 should look at where these systems work and where they
17 don't work. To Alondra's point, to Ms. Nelson's
18 point, the -- there are limitations based on the way
19 these systems are trained. But the important thing
20 with all of the technologies that we deploy is that we
21 understand the four corners of the box in which they
22 work and don't try and use them in the off label

1 methods for which they were used.

2 The early issues with law enforcement using
3 facial recognition were, in my view, similar to off
4 label use of medication. They took something that was
5 built for one purpose, they didn't understand the
6 limitations and they immediately deployed in another
7 purpose and it didn't work. And we should not do
8 that. We should understand how we use those sorts of
9 things. But facial recognition and biometric
10 recognition are some of the most important tools that
11 we have for identifying and keeping track of bad
12 actors, frankly. And that's, you know, and that --
13 and we need to continue to explore, but we need to do
14 it in a way where we are always asking the question
15 about, you know, where it works and where it doesn't
16 work and what the risks are.

17 It really is a different domain from law
18 enforcement. And we need to keep in mind that the
19 rules of engagement are different. And it is not a
20 civilian use of the technology. It's a use for
21 keeping the nation safe from the worst actors on the
22 planet that want to do us harm.

1 MS. WILLIAMS: Thank you. My final question
2 is for Ms. Tabassi. You know, one of the concerns is
3 that the AI wont be accurate, right? And you see that
4 kind of at a very basic non AI level with credit
5 reporting. One reason that people are now allowed to
6 request their credit reports and look at them is to --
7 so that they can look for inaccuracies so that -- to
8 ensure that the data is correct for which decisions
9 are being made.

10 From a NIST perspective, do you have any
11 recommendations for ways to increase the accuracy?
12 Are there ways in a national security context that
13 would allow people to confirm that their information
14 is correct if it's being used?

15 MS. TABASSI: Right. Thank you so very much
16 for that question. You're pointing out to the
17 important topic of evaluations and being able to
18 measure. First, we need to know what that accuracy
19 is. And for a lot of these systems, we don't know.
20 We have the anecdotes and experiences that they
21 hallucinate it and give the right answers and so many
22 other risks, but we don't quite know how to measure

1 accuracy, the false positive, false negatives? Are
2 they still applicable or not? Or do we need a
3 different metrics?

4 And also, measuring just for the accuracy in
5 the laboratory setting doesn't gives us a good
6 understanding and wholesome understanding of all the
7 risks, harms and impacts that can happen in the actual
8 context of use. So, from this NIST point of view, we
9 emphasis a lot on the measurement and we -- not only
10 measurement in the laboratory setting, but also
11 measurement in the actual native real world context of
12 the use of the algorithm.

13 We also know that all of this -- the science
14 of AI evaluations is at a nascent stage. And that's
15 where we need to put a lot more research and
16 understanding on how to do this. Hope that was
17 helpful.

18 MS. WILLIAMS: Thank you. And with that,
19 I'll turn it back to Sharon.

20 MS. FRANKLIN: Thank you. Okay. So, we're
21 going to try and have a more lightning round this time
22 working through each board member getting a chance, I

1 think for just one question. So, my question is to
2 both Alondra Nelson and Elham Tabassi, having worked
3 on the OSTP and the NIST frameworks in this space.
4 And multiple people have spoken already about how, you
5 know, national security, of course, raises unique
6 considerations.

7 And Dr. Nelson, you mentioned a carve out
8 even in the blueprint that you worked on for national
9 security. But I'm wondering if you can point toward
10 with -- when you do think about safeguards that can be
11 put in place to protect privacy and civil liberties in
12 addition to, of course, the basic safeguard of having
13 robust oversight. Are there any particular types of
14 safeguards that you would point to beyond oversight in
15 general that you think are or can be particularly
16 effective in the national security space understanding
17 the particular considerations that are involved in
18 that context that are different from other uses of AI?
19 So, maybe we can start with Dr. Nelson, and then move
20 on to Elham Tabassi.

21 MS. NELSON: Yeah. Just briefly, I think it
22 Senator Rounds, you know, one, they're kind of talking

1 about metrics. I mean, are the tools, the use of the
2 tools doing, you know, fulfilling the mission or not
3 and how do we, you know, collect that data and analyze
4 that data? I mean, that remains the, I think the key
5 way to answer that question. Obviously internal to
6 the IC, we're getting better and more robust tools at
7 doing auditing of systems both, you know, before you
8 deploy them and after.

9 And I think part of what's been encouraging
10 about the last couple of years in the space of AI
11 governance and evaluation more generally is that we're
12 starting to see an ecosystem of different kinds of
13 auditing, red-teaming, both adversarial and otherwise
14 kinds of tools that allow us to know a bit more about
15 the two -- about how the systems work.

16 MS. TABASSI: Thank you for the question.
17 Very quickly. So, it's important to test. My first
18 answer is test, test, test at all of these stages of
19 the lifecycle. But also, it's important to get a lot
20 of these considerations into the design of the system.
21 So, instead of just wait until later and then test the
22 system to see if it's private enough or not, what are

1 the things that -- what are the mechanisms and
2 techniques that can be implemented and designed into
3 the tools, the technology to make it, for example, all
4 of the work around the privacy enhancing technologies
5 to build the technologies, their models that are
6 inherently more secure, more private?

7 So, both at the time of the design and
8 development and do more testing across the whole AI
9 stack or lifestyle. Thank you.

10 MS. FRANKLIN: Thank you. Over to Ed Felten.

11 MR. FELTEN: Thanks. I have a question for
12 the other two panelists. Mr. Souleles and Mr. Usher,
13 based on your experience working in intelligence
14 agencies, the conversation about AI privacy and civil
15 liberties is often framed as a kind of reactive story
16 that AI comes along, it erodes privacy and civil
17 liberties and we look for policy interventions to
18 minimize the damage.

19 Well, my question is about how we might flip
20 that script. Are there proactive ways to use AI
21 within government to strengthen privacy and civil
22 liberties and to reduce other kinds of risks? And I'd

1 like to ask the question both in general and also
2 specifically, what should government agencies be doing
3 toward that goal? Mr. Souleles first, please.

4 MR. SOULELES: Yeah. I think I would start
5 by, again, going back to definitions, what do we mean
6 by bias and systems, right? Often, when we have a
7 conversation with privacy and civil liberties folks,
8 bias means we're denying or causing some harm to some
9 class of individuals based on the fact that the data
10 itself is biased.

11 Computer systems don't have that concept.
12 And data analytic systems don't have that concept.
13 All data analytic systems are biased. We make biased
14 decisions when we choose what data to include and what
15 we choose what data not to include. And we make
16 biased decisions when we decide what questions to ask
17 or what not to ask. So, when we ask, can we eliminate
18 bias in our systems? It's actually a false question.
19 You cannot because all data analytic systems are
20 biased. There's neither good bias nor bad bias.
21 There just is bias in the systems.

22 The important thing is to understand the

1 biases and deploy them where they -- and only deploy
2 the systems where we know they are workable. And
3 that's complicated and not always obvious. So, I
4 would say that what the community can do is to begin
5 to develop more and more and better data analytic
6 tools to describe the biases in the data that it
7 already has and to make sure that we set appropriate
8 guidelines around the use of that data.

9 MR. FELTEN: Mr. Usher?

10 MR. USHER: Sure. I'll just build on what
11 Dean was saying and actually allude to a point that
12 the Dean made earlier in his remarks that, you know,
13 the machine learning systems applied against a rule
14 can be fairly effective. So, you asked to flip the
15 script. The IC right now has a set of guidelines and
16 rules that it follows with regard to the use of U.S.
17 person's data. And humans operating today in the
18 Intelligence Community sometimes make mistakes.

19 And they put wittingly or unintentionally
20 such data into a report or an assessment or something
21 like that. And it takes other humans to catch the
22 error and to remove that information. One could

1 imagine that an AI-enabled tool would be much more
2 efficient and fast and would have perhaps a greater
3 scope of reach across everything that the, say the
4 National Security Agency is producing to make sure
5 that it complies with established guidelines with
6 regard to privacy protections or other guidelines with
7 regard to quality, transparency, application of
8 tradecraft and proper classification.

9 MR. FELTEN: Thanks. Onto Travis LeBlanc.

10 MR. LeBLANC: Thank you, Ed. I have a
11 follow-up question to Mr. Souleles. I completely
12 agree with you that when it comes to data sets and the
13 use of AI that bias is likely to be -- bias is
14 inherent, not likely to be inherited. It will always
15 exist. Where I do digress from your view is I do
16 believe that there is bad bias that is out there. And
17 whether you agree with that or not, it's apparent that
18 someone is deciding what bias is acceptable when
19 deploying an AI system.

20 How can we ensure that that decision-making
21 is more transparent even in the national security
22 context so that at least the public or other decision

1 makers can know that a particular calculus or
2 acceptance of bias was being made? And you're on
3 mute.

4 MR. SOULELES: Sorry. I think it's important
5 that we require that our deployment of systems that
6 our data-based and are trained on data that is
7 collected from any source, that we have pretty strict
8 guidelines on how we analyze that data and classify it
9 and determine that and that we have policies and
10 procedures in place to actually assess the biases in
11 the data.

12 For example, and by the way, I don't disagree
13 with you. I agree that there is bad bias in our data.
14 If you were to train a system to make loan decisions
15 based on loans that were made in the 1950s, in the
16 south, you would get a very biased system and you
17 would reinforce that bias.

18 That is not the kind of bias that any of us
19 want to see reinforced. That's why I say it's really,
20 really important for us to understand the data that is
21 being put into the system. And there's no magic
22 bullet here. It requires people with data science and

1 analytic skills. It requires social science skills.
2 It requires a whole broad range of skill so that we
3 even -- sometimes we don't even know the question to
4 ask. And if we don't know the question to ask, we're
5 never going to know what the -- and we may not know it
6 until it produces a bad result. And that is
7 problematic.

8 But as SCSP mentioned earlier, I don't think
9 it's a reason to stop. I think it is a reason to
10 continue to ask the question and ensure that our
11 agencies are actually doing the things that they say
12 they are doing. You know, I sent out a summary of the
13 U.S. intelligence community's privacy and some of
14 these (phonetic) guidelines, and they do -- they say
15 everything you want them to say, right? The question
16 is, are they actually doing it? And do we have enough
17 oversight to make sure that they're doing the things
18 that they say they are doing?

19 MR. LeBLANC: Thank you very much. And I
20 will pass it on to Beth Williams.

21 MS. WILLIAMS: Thank you very much. So, for
22 a final question, you know, I thought I would turn it

1 all over to you to get your thoughts on this. When we
2 discuss AI and I think in the general discussion it's
3 often very esoteric, it gets to very high level
4 principles and many people don't understand what it
5 actually means to use AI in a national security
6 context.

7 And so, I'm wondering if you can share your
8 view of perhaps the most promising use of AI in a
9 counterterrorism situation. If all of you -- if you
10 have examples that you could share, that would maybe
11 put some meat or explanation to what this actually is.
12 And we can start with Dr. Nelson.

13 MS. NELSON: So, I would go back to my DART
14 mission example. I mean, it's intended to be about an
15 asteroid. But you could imagine that technology,
16 that's mean that shifts the trajectory of something
17 that's coming towards, you know, the United States or
18 the planet could be used for, you know, weapons and
19 these sorts of, you know, kind of spatial warfare.
20 So, I think that's -- I'm a big fan of that one.

21 MS. WILLIAMS: Okay. And Ms. Tabassi, so,
22 our board, actually it's supposed to be looking at

1 counterterrorism applications and we can look at
2 programs that also touch on counterterrorism. But do
3 you have any ideas with regard to how it can be used
4 specifically for counterterrorism?

5 MS. TABASSI: I think everything that AI is
6 good for and that is trying to understand and analyze
7 a lot of data and make -- improve the data analytics.
8 I cannot think of a particular example, but anything I
9 have found often of what Alondra just said about the
10 (inaudible).

11 MS. WILLIAMS: Okay. And Mr. Souleles?

12 MR. SOULELES: Let's see here. Yeah. So,
13 imagine that you are a young analyst working in the
14 National Counterterrorism Center at the Office of
15 Director of National Intelligence and your job is to
16 come in, in the morning and read your (inaudible) of
17 all of the reporting that's happened overnight. And
18 there may be many thousands of reports that have flown
19 in from all around the world, both open source and
20 classified.

21 And your job as the first order, first
22 guideline is to sort through all that and find out

1 which of those things might be important to the
2 question of the day. And the question of the day is a
3 different question today than it was yesterday, right?
4 The difference between September 10, 2001 and
5 September 11, 2001, you're asked to analyze a new
6 question today.

7 And your job -- and you're the most junior
8 analyst in the department and your job is to be on
9 that watch and just pick out the things -- you know,
10 we call it swipe left and swipe right, you know, for
11 reporting. And your job is just to pick the things
12 that are most useful for the next level up to actually
13 read and recognize. And you have maybe two or three
14 seconds to look at each report before you make that
15 decision.

16 That's an area where computer analysis,
17 summarization, all the kinds of things that these
18 things we know do actually really, really well and
19 they're not making any assessments or judgments,
20 they're just saying, let's create a sieve so that the
21 human analyst gets to look at the most important
22 things and not the least important things." So, I

1 think that's a very specific recommendation for the
2 kind of things that we could use today and would be
3 actually of great benefit.

4 MS. WILLIAMS: Thank you. Mr. Usher.

5 MR. USHER: I'll give you a real-world
6 example and we don't have to go too far back in
7 history to find it. But in 2018, Israel's
8 intelligence service, the Mossad, according to press
9 reports, sent a team into Iran and secretly raided a
10 vault that contained the nuclear archive for Iran's
11 nuclear program. The team sat on the ground for
12 several hours and stole about 20 percent of that
13 archive.

14 And according to the press accounts, that
15 included about 55,000 documents and about 55,000 CD-
16 ROMs with audio and video files, almost all of which
17 was in Farsi. And they brought that back to Israel
18 for exploitation. And you can imagine the pressure
19 that was on the Mossad analysts who were charged with
20 taking that raw data and trying to make sense of it to
21 answer the urgent question at the time as to whether
22 or not Iran's nuclear program, which had existed from

1 1999 to 2003, was in fact continuing, or perhaps the
2 world misunderstood where they left off in their
3 capabilities.

4 The Israeli team charged with making sense of
5 that vast amount of data took months to process that
6 information. With today's AI capabilities, and I'm
7 talking here broadly about even earlier versions of AI
8 such as machine translation, which is not quite
9 effective, the first two steps in that analytic
10 process, translating the material into Hebrew or other
11 languages, and identifying salient points within that
12 data that the analysts should look at and in which
13 priority can happen now within minutes, if not
14 seconds, right?

15 This is a tremendous advantage when dealing
16 with the intelligence challenges of the future, where
17 we'll be looking at large datasets, entire computer
18 networks, or a foreign country's AI stack, where it's
19 impossible for humans, even large teams of humans, to
20 go through that accurately, reliably, quickly. AI is
21 a real boon to the intelligence community in a
22 situation like that.

1 MS. FRANKLIN: Thank you very much. Okay.
2 So, that is going to bring our first panel to a close.
3 I'd like to thank each of our panelists for this first
4 panel, for sharing your insights with us. And for our
5 audience, we are now going to take a short 5-minute
6 break and we will then return for our second panel.
7 Thank you

8 (Recess)

9 MS. FRANKLIN: Okay. Thank you. We are now
10 back for our second panel. And I'd like to welcome
11 them all here. We will again, with this panel,
12 proceed through the panelists in alphabetical order
13 for brief opening statements, and then move on to
14 board member questions. This time the board members
15 will reverse the order of board member questioning.

16 So, our panelists for this panel are Miranda
17 Bogen, who is director of the AI Governance Lab at the
18 Center for Democracy & Technology; Clare Garvie, who
19 is counsel at the National Association of Criminal
20 Defense Lawyers; Jamil Jaffer, who is director of the
21 National Security Institute at George Mason Law
22 School; and Peter Winn who is acting chief privacy and

1 civil liberties officer at the Department of Justice.

2 So, first to Miranda Bogen, for your opening
3 remarks.

4 MS. BOGEN: Thank you so much. And thank you
5 to the Privacy and Civil Liberties Oversight Board for
6 the opportunity to provide comments today about the
7 privacy and civil liberties implications of AI. My
8 name is Miranda Bogen, as mentioned, and I'm the
9 director of the AI Governance Lab at the Center for
10 Democracy & Technology, which is a nonprofit and
11 nonpartisan organization that defends civil rights and
12 civil liberties and democratic values in the digital
13 age. The AI Governance Lab works to develop
14 actionable and practical efforts to govern AI -- the
15 use and development of AI responsibly.

16 Prior to joining CDT, I worked with
17 developers and deployers of advanced AI systems and
18 machine learning models at Meta, where I was directly
19 involved in defining processes for managing risks
20 presented by these technologies, and building
21 approaches and guidance to encourage the adoption of
22 more responsible AI development practices.

1 The newest AI powered methods and tools can
2 offer benefits for organizations and government
3 actors. But we urge caution, especially when
4 considering uses in high stakes contexts such as
5 national security and counterterrorism, given the many
6 well-known, but unresolved risks that AI systems pose
7 to people's rights and safety.

8 First, intelligence agencies may seek to use
9 AI to help analyze and act on huge swathes of text,
10 audio, image, and video intelligence. We're deeply
11 concerned, however, that without appropriate
12 safeguards and oversight, this technology will be
13 deployed to facilitate and dramatically expand
14 indiscriminate surveillance and increased reliance on
15 automated tools to inform national security
16 activities. Incomplete, unrepresentative, and biased
17 training data can lead to erroneous discriminatory and
18 harmful outcomes, and even functional AI tools can
19 lead to the suppression of dissent and the oppression
20 of marginalized groups.

21 In addition to embedding pernicious biases
22 that may be challenging to detect, in many cases, AI

1 outputs are highly arbitrary because the process of
2 training machine learning and AI models unavoidably
3 involves a significant amount of randomness, which
4 risks leading to erroneous outcomes that will
5 disadvantage and harm people.

6 Second, AI powered systems remain inherently
7 unreliable and difficult to scrutinize, making
8 oversight critically important. Simply put, the
9 intelligence community should not assume that AI
10 augmented analysis is by default more accurate than
11 human analysis. AI systems remain vulnerable to
12 subjective judgments reflected in training data, as
13 well as to the human interpretation of outputs,
14 hallucinations, and changes the system settings that
15 can lead to increased errors and flawed outcomes.

16 To maintain some degree of confidence in the
17 performance of an AI system, independent oversight
18 should involve making sure that agency's focus on
19 training data used to develop AI systems was lawfully
20 and ethically gathered, and is relevant to the
21 system's intended uses. Supporting transparency into
22 how systems are customized, fine-tuned, and validated

1 for national security purposes, and maintaining
2 visibility into how these systems are integrated into
3 operational work and how their outputs are acted on in
4 order to prevent the erosion of safeguards against
5 errors and biases.

6 Third, ensuring human decision makers with
7 subject matter and domain expertise can and do
8 maintain meaningful oversight over the use of AI
9 systems, will require proactive effort. National
10 security institutions must put in place internal as
11 well as independent governance mechanisms to promote
12 the responsible use of AI. They should clearly assign
13 decision making and internal oversight
14 responsibilities, require review and approval by high
15 level officials for the procurement of systems and
16 scrutiny of use cases that present particularly high
17 risk. Privacy, civil liberties, and legal officials
18 should be given comprehensive visibility into how
19 departments and agencies are using AI and must be
20 included as part of the decision-making process
21 through the AI development, procurement, and
22 deployment lifecycle.

1 Fourth, AI should not circumvent rules and
2 safeguards established for intelligence agencies and
3 personnel. For example, if Congress requires court
4 approval before the results of U.S. person queries, a
5 702 collected communications can be reviewed.
6 Intelligence personnel might seek to use AI to
7 circumvent such a rule by tasking an AI system to
8 review the communication based on the position that no
9 human review was conducted and thus no court approval
10 was required. Things like this should not be
11 permitted.

12 Finally, PCLOB should assess compliance with
13 insufficiency of existing executive policies on
14 agency's use of AI. As an independent oversight
15 agency with access to classified programs, you are
16 uniquely poised to assess the effectiveness of
17 administration policy on agency's use of AI, including
18 ensuring that the forthcoming memorandum on national
19 security uses of AI is applied narrowly, only to those
20 uses of AI exclusively centered on national security.
21 Other AI applications are subject to the OMB
22 governance memorandum.

1 As intelligence and national security
2 agencies deepen their pursuit and investment in
3 technologies like artificial intelligence, the careful
4 consideration of privacy and civil liberties
5 implications of AI systems is both necessary and
6 urgent. Independent oversight and expertise will play
7 a critical role in ensuring that decisions around the
8 appropriate use of AI power tools remain grounded in
9 human rights and core democratic values.

10 Thank you.

11 MS. FRANKLIN: Thank you. Next, Clare
12 Garvie.

13 MS. GARVIE: Thank you so much for inviting
14 me to speak on this panel with you today.

15 I want to start with an example because I
16 think it's a helpful illustration. So, 10 years ago,
17 an Israeli company called Faception began marketing an
18 AI based system to identify possible future terrorists
19 in real time, without any prior intelligence of the
20 person required. The tool, according to the startup,
21 could predict someone's propensity to be involved in
22 future acts of violence, based on an analysis of their

1 facial features, captured in video at a distance.

2 When asked by a Wall Street Journal reporter
3 back in 2018 about the foundational validity or
4 reliability underpinning the tool, Shai Gilboa, co-
5 founder and CEO of the Faception stated, "I need to
6 emphasize that there is no scientific evidence for the
7 terrorist classifier." Nevertheless, this system
8 continues to be promoted, and is used by at least two,
9 as of yet, unnamed country's defense agencies. The
10 company also markets tools to identify possible white-
11 collar criminals, pedophiles, brand promoters, bingo
12 players, and academic researchers.

13 I highlight this tool not because I suspect
14 that U.S. is one of the countries using it. I have no
15 evidence one way or the other. But because I think it
16 illustrates many of the privacy, civil liberties,
17 reliability, transparency, and other concerns with AI
18 that we're here to discuss today. And we've already
19 heard a fair amount about including; one, the often-
20 unquestioned impulse to see AI as providing a solution
21 to all intelligence, national security, or law
22 enforcement challenges. This ability to identify the

1 next potential plot, screen travelers, gather
2 evidence, without necessarily considering the true
3 costs or evaluating viable alternatives.

4 Two, closely related, the fact that AI may
5 over-promise and under-deliver, put simply, we risk
6 deploying junk science in an extremely high
7 consequence environment, both on the national security
8 side and for the people investigated or denied access
9 or benefits based on AI determinations.

10 Three, the threat of entrenching existing and
11 often biased heuristics about who or what constitutes
12 a threat. Faception's terrorist classifier appears to
13 look for Middle Eastern male faces. It failed to flag
14 Ted Kaczynski as a possible threat for example, and at
15 least initially, it was not trained on women at all.
16 This bias is well documented across facial recognition
17 deployments, but is in no way unique to facial
18 recognition systems alone.

19 Four, the increased reliance on AI to define
20 and identify what constitutes anomalous and often
21 suspicious or probable cause level behaviors or people
22 risking supplanting human and judicial determinations

1 of probable cause, and in some cases, even guilt.

2 And five, exacerbating the others, a tendency
3 for AI systems to add layers of opacity onto already
4 deeply non-transparent sectors, like intelligence and
5 national security.

6 To narrow this focus somewhat, as this board
7 is of course acutely aware, two of the core mechanisms
8 to ensure privacy and civil liberties in the
9 intelligence and national security space are; one, the
10 minimization of collection, retention, and
11 dissemination of U.S. persons' data. And two,
12 transparency and oversight.

13 In evaluating national security applications
14 of artificial intelligence, I urge the board to
15 consider that AI and the promise that many of its
16 applications hold out, is in tension with these
17 mechanisms. Many AI systems brought the ability to
18 ingest and make sense of vast quantities of disparate
19 information about people, associations, behaviors, and
20 more. This combined with system needs for large
21 representative training datasets, creates an incentive
22 for more, not less data collection, retention, and

1 dissemination.

2 On the transparency and oversight mechanism,
3 the black-box nature of algorithms coupled with trade
4 secret claims that accompany private sector
5 development of algorithms often leave agency users
6 themselves, not to mention the public, uninformed
7 about potential sources of error and bias and threats
8 to privacy and civil liberties. This is exacerbated
9 by the rapidly evolving nature of AI based systems, a
10 pace that I believe our current structure of privacy
11 impact assessments, systems of records notices, and
12 other transparency mechanisms have little hope of
13 keeping up with.

14 I further suggest first and foremost,
15 orienting to the question of whether is a tool
16 necessary? And if it is, is it necessary that that
17 given tool be AI based at all? Or does the data
18 collection transparency, reliability, and bias
19 concerns posed by the system and introduced by the AI
20 component outweigh the purported benefits? I also
21 encourage the board to push executive agencies to
22 think critically about whether the current oversight

1 and transparency structure is adequately responsive to
2 the realities of AI, its pace of development and
3 deployment in the face of those harms.

4 Thank you so much. I look forward to
5 answering your questions.

6 MS. FRANKLIN: Thank you. We'll next hear
7 from Jamil Jaffer.

8 MR. JAFFER: Thank you, Chair Franklin, and
9 board members for having me here today. My name is
10 Jamil Jaffer, I'm the founder and executive director
11 of the National Security Institute at George Mason
12 University's Antonin Scalia Law School. I'm thrilled
13 to be here today at this forum as PCLOB takes
14 advantage of its statutory responsibility to take
15 action, analyze reactions of executive branch that are
16 focused on protecting the nation from terrorism.

17 Today, the threat of terrorism is extreme.
18 We just heard in the last few weeks from the FBI
19 director that he believes that he is hard pressed to
20 think of a time at which so many different threats to
21 our public safety and national security were so
22 elevated all at once. We know the world is on fire.

1 We see the wars in Ukraine, the war in the Middle
2 East, a potential threat from China in the Indo-
3 Pacific. And the FBI director is telling us that the
4 threat from foreign terrorists has risen to a whole
5 another level since the October 7, 2023, terrorist
6 attacks on Israel by Hamas.

7 Director Wray went on to note that there's
8 already a heightened risk of violence in the United
9 States before October 7. And since then, the FBI has
10 seen a rose gallery of foreign terrorist organizations
11 call for attacks on Americans and their allies,
12 raising concerns. Not only that individuals and small
13 groups will draw twisted inspiration from what's
14 happened in the Middle East, but there's increasing
15 concern by the potential for a coordinated attack here
16 in the homeland. A (Inaudible) attack conducted in
17 Moscow by ISIS K, ISIS Khorasan, that took the lives
18 of over 150 or nearly 150 and injured over 500. The
19 threat is extreme.

20 In fact, Graham Allison and former deputy CIA
21 director, Michael Morell, reported in foreign affairs
22 just last month, that the terrorism warning lights are

1 blinking red. The United States faces a serious
2 threat of terrorism in the months ahead. This is an
3 extreme situation. This is not a time to step
4 cautiously and pause on our questions about whether we
5 should take advantage of the AI revolution to counter
6 terrorist threats. Today is a time where we must lean
7 forward. Now, we must do so in the context of our
8 values and the protection and privacy -- and
9 protection of the privacy and civil liberties of
10 Americans. That is critical.

11 But the way to do that is to not slow down
12 what we implement, to not think hard about
13 (inaudible), but to lean forward and to think about
14 how we can build AI capabilities for the national
15 security community, for the counterterrorism community
16 in a way that bakes trust, safety, and security in
17 from the jump at development, in deployment, and on a
18 going forward basis. That doesn't require going slow,
19 but it does require thinking hard about trust, safety,
20 and security.

21 So, how do we do that? How do we bake in
22 trust, safety, and security right from the jump?

1 Well, we're not writing on a blank slate. Luckily, we
2 have a long-time scenario of dealing with these
3 questions in other domains; cybersecurity,
4 counterterrorism, and other domains where we bake
5 trust, safety, and security and at the outset, we need
6 to do more, we need to get better. But the
7 government's already doing this. DHS has secure by
8 design principles, resilience by design principles for
9 software. NIST, as you heard earlier today, has
10 reliable AI standards. NIST has reliable
11 cybersecurity standards. A lot of these standards are
12 built on what industry is doing already and how
13 industry might lean forward.

14 The government can incentivize the adoption
15 of safety, trust, and security in their systems by
16 using their buying power. The government can provide
17 incentives in the form of tax relief, they write
18 incentives in the form of liability and regulatory
19 relief. The government can provide incentives in the
20 form of grants to companies and organizations that are
21 building these capabilities to make them more trusted,
22 to make them more safe, and to make them more secure.

1 And in fact, investors and innovators have an
2 incentive for baking trust, safety, and security into
3 their systems. It makes the products that they build
4 more likely to be adopted by the government. And by
5 industry if they're trusted, safe, and secure. This
6 idea that we need to treat AI, like it's a global
7 pandemic or like it's a nuclear weapon, as some have
8 suggested, is simply wrongheaded. AI has the power to
9 be transformative, we ought to take advantage of it,
10 particularly at this heightened threat level.

11 And just to demonstrate that, in fact,
12 investors and innovators have the incentive to invest
13 in this, the venture capital firm that I work with,
14 Paladin Capital, led a group recently of a dozen
15 venture capital investors, along with the NATO
16 Innovation Fund, signing a series of principles around
17 investment in trust, safety, and security. There's a
18 growing market in this space. This is not a time to
19 go slow. It's a time to lean forward, but to do so in
20 a way consistent with our values and the protection of
21 the privacy and civil liberties of Americans.

22 Thank you for your time, and I look forward

1 to your questions.

2 MS. FRANKLIN: Thank you. And now we'll hear
3 from Peter Winn.

4 MR. WINN: Thank you, Chair Franklin. And
5 thank you to the other members of the board. I look
6 forward to your questions.

7 Before I begin, I just wanted to say
8 something that I think I've drawn from comments that
9 some of the other thoughtful commenters have made,
10 which is AI is a tool and it's used by humans. We
11 have a lot of laws out there that apply to humans.
12 But it's not as if those laws cease to apply when
13 you're using AI. Those laws still apply. If a law
14 forbids discrimination in certain ways, the use of AI
15 to discriminate will violate that law. If an AI -- I
16 mean, there's a lot of examples. My favorite might be
17 the recent example of an AI program that used an
18 actress' voice. Well, AI didn't make the rules about
19 inappropriate appropriation of a person's identity,
20 without their permission, go away just because you're
21 using an AI program.

22 So, what I'm getting at is that the

1 Department of Justice, if we, you know, we collect a
2 lot of information, and we have to use new
3 technologies in order to keep the public safe and
4 protect national security. If we lose the trust of
5 the public when we're doing that, we're going to lose
6 the authorities that we depend on to collect that
7 information that we need to protect people. So, trust
8 is mission critical. And the best way to lose trust
9 is not to comply with the laws that apply to us or not
10 comply with the frameworks that we've adopted.

11 Now, I'd like to spend most of my time
12 talking about the recent executive order on AI.
13 There's been a mention of the national security
14 memorandum on AI that's part of that executive order.
15 I'm not in a position to discuss that because that's
16 still being deliberated. But I would encourage the
17 board to refer back to the 2020 AI framework for the
18 intelligence community that was developed, I guess,
19 now nearly 4 years ago, and how thoughtful and forward
20 leaning that framework is, and how so many of the
21 frameworks concepts that were developed at that time
22 have been now, even see them in the executive order on

1 the safe, secure, and trustworthy development and use
2 of AI.

3 The guidelines and practices aligned with the
4 NIST AI risk management framework are extraordinarily
5 helpful tools in the development of AI. The efforts
6 to mitigate the risks of inappropriate algorithmic
7 discrimination that can may be exacerbated by AI. As
8 some of the commenters pointed out, you can't
9 eliminate bias, you're trying to mitigate bias that
10 you don't want to have happened, inappropriate bias or
11 unwanted bias. All systems are going to be biased.

12 The OMB directives that have been issued
13 implementing the executive order are extremely
14 helpful. So, the Department of Justice has so far
15 designated Jonathan Mayer as our chief AI officer.
16 We've launched the Emerging Technologies Board. And
17 we've complied with our AI use inventory. And it's up
18 on the department's Open Data website. We're looking
19 to include AI assessments as part of the system of
20 procurement and development process. And we encourage
21 the board to review the NIST AI risk management
22 framework and playbook. They're not prescriptive, but

1 those NIST tools, we have found, represent a really
2 excellent roadmap for any organization wishing to
3 engage in conscientious implementation of this new
4 technology.

5 In April, NIST released a draft publication
6 to help manage the risk of generative AI. And the
7 generative AI profile can help organizations identify
8 unique risks posed by generative AI and to mitigate
9 those risks in a way that aligns with that
10 organization's goals and priorities. That profile
11 identifies a group of 12 risks relating to generative
12 AI. Three of those, I think are key, having to do
13 with data privacy, information security, and general
14 information governance. The NIST framework provides a
15 set of actions to help organizations identify,
16 measure, map, and manage those risks consistent with
17 that risk management framework.

18 AI is a novel, emerging technology, but its
19 use cases are generally understandable. And the
20 existing technology neutral legal structures, the
21 government is already subject to, are excellent ways
22 in which we need to implement the AI just as we've

1 navigated other prior technological advancements.

2 For example, the department is required under
3 the E-Government Act of 2002, to conduct privacy
4 impact assessments, whenever it implements a new
5 information technology. Now, the last time I heard,
6 AI was an information technology. So, we're going to
7 be applying our existing sound privacy impact
8 assessment frameworks to the requirements in the EEO
9 to addressing the unique risk factors of AI in a
10 rational and responsible manner.

11 Whenever we implement AI systems, such as
12 facial recognition technologies, we always require a
13 human to be in the loop, where the AI is used to make
14 determinations about individuals. This is a
15 longstanding standard within the department's policy
16 development and practice. And we fully intend to be
17 implementing that basic requirement of having humans
18 in the loop when we're implementing AI programs.

19 I think Dean had a really insightful
20 observation that the usefulness of AI is a
21 relationship between machines and humans. And it's
22 governed by rules. When Deep Blue defeated Garry

1 Kasparov, that was not a defeat that took place
2 because the machine was better than the human or the
3 human running the machine was better than the human.
4 But because the interface and the rules governing that
5 interface, that was what made the difference, that
6 made the difference that provided the human machine
7 interface such a powerful tool that it defeated the
8 world's grandmaster. And I think keeping focused on
9 the human relationship to the AI programs that we're
10 going to be implementing is the key to advancement of,
11 you know, those technology.

12 Thank you.

13 MS. FRANKLIN: Thank you. So, we'll now
14 start with questions by Board Member Beth Williams.

15 MS. WILLIAMS: Okay. Thank you very much.
16 And thank you to all of our panelists for being here
17 today. We really appreciate your views and your
18 expertise on these questions.

19 So, my first question is actually for Mr.
20 Jaffer. You talked about the trust, safety, and
21 security. And, you know, focusing specifically on
22 trust, I always think that one of the issues with AI

1 is kind of a confidence problem, right? The only
2 thing worse than somebody having the wrong answer is
3 the guy who's also very confident that his answer is
4 right. And I think that could be a problem with AI in
5 that we, you know, if people believe it to be 99.999
6 percent accurate all the time, they're putting trust
7 in the answers that it's giving them or the outputs
8 that it's giving them.

9 So, my question to you is, are there ways
10 that you've thought about that we could address that
11 confidence issue? Is there a way to put a maybe next
12 to your answer a confidence estimate? Or are there
13 like other AI programs that should be labeled, layered
14 on top of existing AI programs to give human users
15 analyses of how likely to be correct certain outputs
16 are?

17 MR. JAFFER: Yeah, it's a great question,
18 Member Williams. You know, the -- I think part of the
19 challenge with when you talk about AI and its
20 capabilities is, we heard earlier about sort of idea
21 that we're sort of associating human values with AI
22 because it sounds and feels colloquial. So, we trust

1 it the way we trust a human, I mean, trust that it's
2 not sort of freelancing, but even humans, you know,
3 make things up, right?

4 In a lot of ways, the way that AI works by
5 associating words with other words that we -- that it
6 puts together may very well be how we interpret things
7 in our brain as well. We're not actually sure when a
8 person tells them, whether they're actually telling us
9 the truth or not. But we judge them based on a lot of
10 other factors. There's got to be ways to do the same
11 with AI.

12 We're not going to get to a point, I don't
13 think, where we're going to eliminate all of the
14 "Hallucination problem." What we can do, however, is
15 create capabilities like you say, that provide
16 confidence assessments that allow AI models to ingest
17 the data from other models and regurgitate what they
18 see is the right answer amongst a variety of them.

19 Sometimes with some AI models, if you look at
20 Google Gemini, you'll see it'll give you three
21 different versions of the same answer to see which one
22 you are more confident in. And if you had confidence

1 metrics associated with that, that might actually give
2 you more to pick from.

3 And in addition, there are now capabilities,
4 some of which venture capital firms like ours are
5 investing in that actually look at AI output and say,
6 are we getting the right thing? Is the model working
7 in the right ways, where you can sort of put your
8 model in and ensure your model is doing what you want
9 it to do? Those aren't going to ever be a 100
10 percent, but you can get better and better over time
11 and that's a way of creating confidence also.

12 At the end of the day, though, I think people
13 have to recognize that this is a tool and the
14 capability, it's not an answer. So, you know, you
15 just had, you know, Peter talk about a human in the
16 loop. We heard about that earlier as well. There's
17 also this notion of a human on the loop, which is to
18 say, there are some automated decisions to be made,
19 but a human can intervene and stop a decision or walk
20 it back if need be.

21 So, there's a variety of ways that we have of
22 humans engaging with AI. But more often than not,

1 what it really is it's not a substitute for human
2 judgment. It can't and shouldn't be. It's simply a
3 supplement to help a human analyst, a human
4 investigator and the like, do their job better,
5 faster, more effectively.

6 MS. WILLIAMS: Thank you. My second question
7 is for Ms. Bogen. So, how, in your view, should we be
8 looking at the privacy implications from AI review of
9 data as opposed to human review?

10 I'm thinking, for example, very popular web-
11 based e-mail programs, famously in the past, right,
12 scanned people's contents of e-mails to -- in order to
13 provide them better ads. I'm told that's not done so
14 much anymore, but it certainly will be done in the
15 future. And I think for many people, they thought,
16 well, if it's a computer doing it as opposed to
17 someone reading my e-mail, then I'm okay with it.

18 So, how do you look at that and, you know, is
19 it less of an issue or greater issue or the same if a
20 computer does it as opposed to if a human does it?

21 MS. BOGEN: Thank you so much. There were
22 some previous comments indicating that, you know,

1 existing laws and expectations should apply whether
2 we're talking about humans or systems. But I think
3 what's important to remember is the way in which the
4 introduction of AI-powered systems or really any
5 digital technology change our understanding of those
6 processes, and whether those changes in process enable
7 the enforcement of those laws or expectations in the
8 same way that we intend.

9 So, for one, making sure that if there are
10 rules around human access to data, is the intent
11 behind those rules being applied in a similar manner
12 to an AI system. But to your deeper point, I think
13 we've moved into a world where the access to data
14 itself is very much not the only question of privacy
15 as we all know, it's about how that data is used and
16 the actions that it informs.

17 And so, to the extent that information is
18 being reviewed by a system that is going to inform an
19 action that could lead to the same type of harm,
20 either invasion of privacy in accessing information,
21 people didn't realize was being accessed or for
22 purposes that are disallowed, or harm that comes to

1 that person by the analysis of that data. It
2 shouldn't matter whether that was by a person
3 reviewing that data or by a system reviewing that
4 data.

5 And so, the oversight can come in thinking
6 about how is this system being actioned? What -- how
7 are the outputs being presented to humans? And how
8 are the outputs leading to actions that are more or
9 less reversible? So, even if you had a human in the
10 loop, are they empowered to do something about a
11 system if it's behaving erroneously?

12 So, that sort of review of precisely what
13 action is a system being instructed to take and how is
14 that action triggering additional action should be the
15 focus of oversight, regardless of where the access to
16 data is coming throughout that process?

17 MS. WILLIAMS: Thank you. So, my next
18 question is to all of our panelists. And it's
19 actually Member Felten's question from the first
20 panel, which I think is a very good one, which is, how
21 do you think AI can be used to actually enhance
22 privacy and civil liberties protections?

1 We always talk about the concerns about using
2 it for other uses. But could you all share if you
3 think, and if so, how AI can be used to improve
4 privacy and civil liberties? And we can go in any
5 order. I see Mr. Jaffer has his hand up, so happy to
6 go with you first, and then then proceed to the other
7 panelists.

8 MR. JAFFER: Well, I already had a chance to
9 talk, but I'll talk very quickly about it, which is to
10 say, I actually think there's a real opportunity here
11 to use AI for privacy-enhancing purposes. If you
12 think about it, programs like the metadata program,
13 which was highly controversial and caused a lot of
14 controversy when it was first disclosed, actually can
15 be very privacy-enhancing in the following way.

16 If in fact what you're going to do to find
17 out whether somebody is a terrorist or not, when you
18 have a suspect number, is going to do a full content
19 collection. If instead you're using metadata to
20 exclude a whole set of numbers from potential content
21 collection where you already have some amount of
22 predication, some amount of probable cause, you can

1 eliminate a whole slew of people you might do very --
2 much more invasive collection on.

3 AI can play that same role by reviewing a
4 large amount of data rapidly and vetting out a bunch
5 of people you might do a lot deeper dive on without
6 having to put human eyes, human hands on that data,
7 that can be a real advantage. In the same way, you
8 know, controversial program drone strikes and the
9 like, right? They've actually allowed us to take much
10 more precision strikes, a lot less casualties of
11 civilians and the like, look, it's not perfect, but
12 there are places where technology advances
13 dramatically, and actually gains us benefits on the
14 morality, the values that we have, our core values.
15 Even though at the outset, they may seem somewhat off
16 putting and scary, turns out when you apply them the
17 right way, bake in, as we talked about trust, safety,
18 and security, you can get a real advantage, actually
19 be privacy-enhancing as long as you're not afraid of
20 them, and slow walk the whole implementation.

21 MS. WILLIAMS: Thank you. Mr. Winn?

22 MR. WINN: Thank you. Thank you, Member

1 Williams. That's a wonderful question. And, yes, we
2 should thank Ed for the -- or Member Felten for the
3 good question.

4 Two things occur to me. One is obviously in
5 connection with cybersecurity, which is an aspect of
6 privacy that we often forget. The threats that are
7 coming into systems from hackers are really getting to
8 the point where the hackers are certainly deploying
9 bots and other technologies that are in position to
10 overwhelm humans that might otherwise try to be
11 protecting those systems.

12 And so, AI has been used effectively to
13 identify and segregate out the threat, you know, the
14 threat attackers as opposed to the legitimate uses of
15 access to systems. So, that's one example where more
16 effective cybersecurity enhances a very critical
17 privacy interest in the data being used as appropriate
18 and not being unauthorized access.

19 But the other thing that occurs to me is that
20 AI can be thought of as -- generative AI programs can
21 be thought of as a mirror of, I mean, they're bringing
22 out what humans are doing. And humans, as we all

1 know, are bias creatures, we have a lot of biases
2 we're not often aware of, a lot of biases we're not
3 particularly proud of, that we're often not aware of.

4 And so, AI can be a mirror that can be a very
5 unflattering mirror, showing us aspects of ourselves
6 that we don't really want to focus on. I think that -
7 - that the sentencing controversies, the use of AI in
8 connection with sentencing or detention decisions have
9 brought out the unpleasant reality of the data that
10 was being trained on was showing up the human biases
11 that were in -- were always there.

12 And so, in many ways, AI can show us aspects
13 of ourselves that are very painful, but also give us
14 great opportunities to improve and learn from
15 ourselves so that the ugly aspects of our characters,
16 the failures can be then viewed as opportunities for
17 continuous improvement.

18 MS. WILLIAMS: Thank you. And Ms. Garvie or
19 Ms. Bogen?

20 MS. GARVIE: Sure, just to build a little bit
21 on something that Mr. Jaffer mentioned and that's this
22 baking in safety security. And I would add civil

1 liberties and civil rights into that as well, is that
2 to the extent that we're building new tools, new AI
3 tools or other tools right now, we have more levers to
4 pull than if we're retroactively looking at already
5 implemented tools.

6 So, from a privacy and civil liberties
7 protection view, we actually have a unique opportunity
8 now in the pre-implementation stage to think really
9 critically about, can we build this stuff in by
10 design, as opposed to can we retroactively try to
11 build policies around already implemented tools?

12 MS. WILLIAMS: Thank you. And, Ms. Bogen, do
13 you think that there are ways that we can use AI
14 proactively to protect privacy and civil liberties?

15 MS. BOGEN: Well, yes, I agree with Mr. Winn.
16 I think AI, the use of AI or any technical tool can
17 make legible decisions that were otherwise subjective
18 or happening informally, and in that way help to --
19 help oversight entities, whether internal or external,
20 identify patterns of potential misuse and correct them
21 as well as to build in specific safeguards into the
22 technology to protect against actions that are

1 otherwise disallowed.

2 There are other potential uses of AI, for
3 example, to identify, you know, to a spot and redact
4 identifiable information and datasets or to prevent
5 the display of certain information to people who don't
6 have access to it. But I would say those are still
7 remain quite unreliable at this point. But there
8 could be possibilities that AI could help play that
9 role.

10 It really depends on what goal an AI system
11 is oriented toward, and often they're oriented toward
12 an outward goal rather than inward ones.

13 MS. WILLIAMS: Thank you.

14 MS. FRANKLIN: I think we're now over to
15 Travis LeBlanc.

16 MR. LeBLANC: Thank you to everyone for
17 joining us today and for giving us a little bit of
18 your afternoon. I have my first question for Mr.
19 Winn. You've mentioned appropriately, in my view,
20 that existing laws apply to artificial intelligence
21 systems. And, you know, generally speaking, there
22 actually aren't exceptions in the laws that say,

1 except when using an AI system. How does Executive
2 Order 12333 apply to the intelligence community's use
3 of artificial intelligence?

4 MR. WINN: Well, 12333, as you know, Member
5 LeBlanc, is a general framework for how the United
6 States engages in its foreign intelligence work and
7 that's generally through that executive order. There
8 are statutes as well that are part of that framework.
9 But it's a general framework.

10 The Section 2.3 of 12333 as you, I'm sure
11 you're well aware, talks about the importance of
12 maintaining privacy protections in connection with
13 that activity. And the attorney general and the
14 director of National Intelligence issue guidelines for
15 the agencies, these are binding guidelines, for all
16 the intelligence agencies that are developed with
17 usually, in the last iteration of the guidelines, I'm
18 proud to say, the PCLOB was involved in the review of
19 those guidelines. I think that what's likely to
20 happen in the next review of those guidelines, and I'm
21 only speaking for myself, is that the more and more
22 artificial intelligence systems are used, we're

1 probably going to see revisions of the guidelines to
2 incorporate some of the wisdom that we've been
3 developing to try to mitigate some of the risks and to
4 continue the process of implementing this technology
5 in a way that's safe and secure, that maintains the
6 trust of the American people.

7 MR. LeBLANC: Okay. A follow-up question
8 related to that, which is what kinds of revisions to
9 the AG guidelines do you believe would be appropriate
10 for AI applications and uses?

11 MR. WINN: I don't want to speculate on that
12 question yet because we haven't started the next round
13 of revisions. But I think the -- when -- first of
14 all, I think the framework that was issued in 2020 by
15 a former PCLOB attorney who has oversaw that process,
16 Ben Huebner. And that framework really represents, I
17 think, a really quite extraordinary forward leaning
18 tool. And I'm -- I would hope that the national
19 security memorandum that's about to come out, will
20 echo many of those concerns.

21 But I think that those types of -- the goal
22 is to use the technology in a way that maintains

1 trust, to get the benefits of the technology and to
2 mitigate the risks. And I think that -- that the more
3 we learn about how best to do that, and AI is sort of
4 making us acutely aware of our ignorance. But staying
5 aware of our ignorance is probably the best insurance
6 policy that we have that we're not going to be
7 deploying the technology in a way that's going to
8 destroy trust.

9 MR. LeBLANC: Thank you, Mr. Winn. My next
10 question is for Ms. Bogen.

11 Ms. Bogen, you identified several governance
12 mechanisms in your opening remark that -- remarks that
13 you believe should be put in place for responsible
14 governance of artificial intelligence. I think much
15 of what you've covered would largely be true of all
16 government agencies or really any organization that is
17 deploying artificial intelligence.

18 Are there any governance measures that you
19 believe are particularly significant or should be used
20 in the national security context?

21 MS. BOGEN: As you mentioned, I think the
22 approaches to governance of AI technology are similar

1 to approaches of governance to organizations in
2 general, complex systems that involve design
3 decisions, value judgments, implementation details, et
4 cetera. So, I would consider what have been the
5 oversight mechanisms that have been effective in
6 spotting potential issues in that regard as a starting
7 point.

8 In general, my experience has demonstrated
9 that there often is a significant amount of low-
10 hanging fruit that in the excitement over the
11 development of new technologies, tends to be its
12 deprioritized relative to its importance, for example,
13 basic documentation of decisions around the design of
14 these systems, such that they can be reviewed and
15 revisited, decisions about what data was used for the
16 training of the system to the extent that that's
17 shared with the government if they're procuring that
18 system, which is a large limitation, details about
19 what tests were run and how they were determined to be
20 relevant to the task that was being assessed.

21 And details about decision -- value judgments
22 that were made in risk management processes, while

1 approaches like the NIST RMF and other mechanisms are
2 fantastic at helping to structure the design and
3 development and review process. They still leave on
4 the table many open questions around how you weigh the
5 information that is revealed throughout that risk
6 management process against the goals of an
7 organization. And those are where value judgments
8 come into play and where sometimes, unfortunately,
9 civil rights and civil liberties end up falling below
10 the line or at least lower than we would like.

11 So, anything that can enable the spotting and
12 review of those types of decisions will support
13 beneficial scrutiny of their development and
14 deployment in the long-term. And they can facilitate
15 the building of other governance mechanisms on top of
16 that. But without that foundation, it will be very
17 difficult to build other effective governance
18 mechanisms.

19 MR. LeBLANC: Thank you. And Ms. Bogen, do
20 you believe that there are any AI applications that
21 should not be used in the national security context?

22 MS. BOGEN: My response, there will be any AI

1 system that is performing a task that we would not
2 want an intelligence mechanism to be doing at all, for
3 example, real time facial recognition, whether that
4 was a human who was very good at recognizing people or
5 an AI system, we wouldn't want pseudoscientific goals,
6 for example, emotion recognition or other things that
7 come to mind.

8 So, being mindful of what are the parameters
9 around which a system is oriented and do those fit in
10 to the overall structure and values of the
11 organization that is deploying the system.

12 MR. LeBLANC: Okay. Thank you. Next, I want
13 to ask a question to Mr. Jaffer. You discussed in
14 substantial detail in your opening remarks about
15 terrorism and in particular the foreign connection to
16 terrorism. But, you know, as I'm sure you're aware,
17 terrorism is also a domestic threat. It's not just a
18 foreign threat. And in fact, domestic terrorism is
19 the number one terrorist threat to the United States,
20 not foreign terrorism, although many may be surprised
21 to learn that.

22 The key privacy and civil liberties challenge

1 in the domestic context is that the government is
2 usually looking at U.S. persons or somehow obtaining
3 through collection, the information from or about U.S.
4 persons. And I fully agree with you that there is a
5 need for the government, if it's going to deploy AI,
6 to build in trust, safety, and security.

7 The fundamental problem that the government
8 and, in particular, the FBI, since you're referring to
9 the FBI director, the fundamental problem they've had
10 in the past is a lack of trust. And so, how can the
11 FBI build trust that its access to massive troves of
12 data about U.S. persons will not be processed through
13 AI systems in ways that are inconsistent with current
14 norms?

15 And are there any limitations or safeguards
16 that you believe should be put in place to protect
17 against AI abuses by the FBI? For example, should the
18 FBI be able to use artificial intelligence to predict
19 who is or may be a criminal?

20 MR. JAFFER: These are great questions,
21 Member LeBlanc. I would say, let me start with at the
22 end of your last question first, which is to say, no,

1 I don't think we want sort of a predictive system
2 predicting who are criminals. It sort of reminds me
3 of that, the movie whose name I'm going to forget,
4 Minority Report. And I don't think anybody's looking
5 to sort of embody a minority report system at the
6 bureau, whether it was highly trusted, which it used
7 to be back in the past or is less trusted today.

8 And as you know, the trust of the FBI has
9 waxed and waned over time back in the post '60s and
10 '70s era when there were the days of the
11 counterintelligence program, Operation CHAOS, the CIA.
12 There was a deep mistrust of the FBI. And we put in
13 place a lot of policies and procedures to address
14 those and bring them back into a more positive light.

15 I think we've seen a decay in that trust in
16 the more recent era as well, in part because of
17 situations that we've seen in both political parties,
18 as well as the popular dimension where there's been a
19 decay in trust in not just the FBI, but all of our law
20 enforcement and rule of law institutions, including
21 the Justice Department. And that's been a real
22 challenge.

1 It's been in part, I think, fomented by
2 overseas actors as well, but there is some -- there
3 are legitimate reasons for some of that distrust. And
4 you referred to some of them, some of the challenges
5 we've seen in programs like 702 and 215, where as a
6 general matter, the FBI has been doing a very good
7 job, but they make errors. They make mistakes. The
8 mistakes sometimes are of large scale.

9 And so, then they self-report these mistakes
10 to the FISA court, and then it turns into this large
11 issue of, look, the FBI is violating privacy and civil
12 liberties, when in fact they're identifying errors
13 they made. Yes, there are mistakes. They're not
14 intentional. The number of intentional violations are
15 very, very few, whether at the FBI or the NSA.

16 And so, what we don't have is an epidemic or
17 a pandemic or any sort of demic of intentional
18 violations of private and civil liberties, but a lot
19 of mistakes and a lot of errors and that erodes trust.
20 You're right to say that. And so, the question then
21 is, how do you rebuild that trust? And that's going
22 to be a challenge. It's going to be a challenge as we

1 deploy tools that are more and more capable, more and
2 more capable of taking in large amounts of data and
3 processing it quickly.

4 I think what we have to understand is the
5 more data you take in, the more data you process, the
6 more likely you are to make mistakes. The question
7 is, what do you do about those mistakes when you make
8 them? Do you put in place policies and procedures as
9 we've done with the bureau, as we've done with other
10 agencies, like the attorney general guidelines, to
11 guide those and to fix those?

12 And how often do you self-report those? How
13 often do you get caught making an error or get caught
14 making an intentional problem? And where there's an
15 intent and there's an actual failure where somebody's
16 done something wrong, do you throw the book at them?

17 We had an example, you know, a famous example
18 of a lawyer who lied to a court to obtain a FISA,
19 right? Changed a material fact. That guy got time
20 served, right? He didn't get time served. He got
21 probation. That is crazy. That guy should have gone
22 to jail for a long time. He should have been stripped

1 of his license. I understand he did lose his license
2 for a while. He got it back. That is unacceptable.

3 When people make failures in the FISA context
4 where you've got ex parte and in-camera proceedings,
5 you have to throw the book at them. Otherwise, that
6 and all the unintentional mistakes get all bottled
7 together. And we have a situation where fundamental
8 trust is undermined.

9 And I think that's an important role that
10 Privacy and Civil Liberties Oversight Board can play
11 when putting out reports like this one on 702 where
12 they're really -- where you're really candid, right,
13 and very clear about intentional versus unintentional
14 mistakes and not sort of combine the two and treat
15 them like they're the same thing because they're not
16 the same thing.

17 MR. LeBLANC: Thank you very much. My time's
18 up. So, I'm going to go ahead and pass the microphone
19 on to Member Felten.

20 MR. FELTEN: Thank you. I'd like to ask a
21 question to all of the panelists related to
22 algorithmic bias. And I'll ask it in the context of

1 facial recognition, which is an area where perhaps we
2 have the best and most extensive data from NIST
3 studies. And in this respect, we see two things
4 happening at the same time. First, we see that
5 according to NIST studies, the demographic
6 differentials in error rate of the very best
7 algorithms are shrinking considerably over time.

8 But on the other hand, we see continued
9 instances of harm to individuals due to, for example,
10 false arrests in a pattern that is very obviously
11 correlated with race. And so, there's some gap
12 between what the algorithms can do and the results
13 that we're getting in the field in this area.

14 So, I'd just like to ask the panelists, you
15 know, if you could talk about what may be happening
16 there and in particular what we might do to address
17 this disparity so that at least we can reduce the
18 level of errors closer to what the algorithms can
19 provide.

20 And let me go through the panelists in
21 alphabetical order starting with Miranda Bogen.

22 MS. BOGEN: Thank you, member Felten. I

1 think when we think about algorithmic bias, there are
2 a number of different lenses through which to consider
3 that. So, one is simply comparative performance of a
4 model itself vis-a-vis its specific goal. Facial
5 recognition, for example, has a very particular
6 mechanism by which to measure if it recognized a
7 specific individual. And by disaggregating that
8 measurement across demographic groups, you can
9 identify if there are those disparities.

10 Running those tests is one way to identify
11 where the gaps are and facilitate attention to closing
12 them. A way to continue making progress in that front
13 is considering what are the groups by which the
14 measurements are disaggregated. Are those salient to
15 the errors that are being made or are there additional
16 disaggregations that would illuminate the causes of
17 those gaps, which may or may not, and likely are, but
18 may not be fully correlated with legally protected
19 groups.

20 So, another approach to conducting these
21 measurements, in addition to disaggregating by
22 predefined groups, is identifying clusters of errors

1 of systems and trying to consider what might be
2 driving those errors by reviewing those errors.
3 That's to address technical bias in that way,
4 disparate performance against a metric.

5 But in, you know, similar systems, there are
6 also questions around, was the goal of the system
7 appropriately defined? Was the target metric
8 reflective of some kind of underlying assumption in
9 the world that incorporates some kind of historical
10 bias against which disaggregating measurements of the
11 system would not reveal and requires considering a
12 system more holistically?

13 And in other cases, systems might reveal
14 biases that are simply reflected in the world where
15 technical intervention are not the most opportune
16 approach to address that, but rather reflecting on
17 overall processes. So, I would divide it in that way.

18 MR. FELTEN: Thanks. Clare Garvie?

19 MS. GARVIE: Thank you for the question. I
20 think I have two points on this. One is the
21 operational conditions point, and that is that NIST,
22 while the tests that NIST performs on face recognition

1 are extremely valuable, they still don't represent
2 what happens in operational conditions. And that is a
3 sociotechnical system, a series of steps for which the
4 algorithm is one of multiple steps.

5 So, until we have actual testing on face
6 recognition in operational conditions, whether that's
7 in the law enforcement or a national security
8 standpoint, I think we are going to have these
9 differences in what the tests show in terms of
10 accuracy, reliability, and bias, and what we're seeing
11 on the ground in practice. For example, the human-in-
12 the-loop, is that a valuable check against
13 misidentification or does the cross-race bias effect
14 actually exacerbate or perpetuate the racial bias in a
15 way that isn't being tested by NIST?

16 And the other point I would raise is that,
17 yes, while the differential error rates across race
18 have declined over the last few years in the top
19 performing algorithms, it does seem that race, sex,
20 and age still impact the accuracy measurements or the
21 reliability scores given to non-mated pairs.

22 And my suspicion here, sorry to get a little

1 wonky on this, is that the algorithms are still
2 confusing class characteristics in individual
3 characteristics, that, yes, algorithms are not looking
4 for race, sex, and age, but they are looking at race,
5 sex, and age to determine individual identity, but
6 those are class characteristics and not individual
7 ones.

8 So, the types of mistakes that an algorithm
9 is going to make is going to be between people who
10 look very similar, aka people of the same race, sex,
11 and age, which again leads us to question whether the
12 human in the loop is actually performing a valuable
13 check, when the algorithm is making the same types of
14 mistakes that humans are going to make, which is
15 confusing people who are in the same demographic
16 cohort.

17 And then we put that all into a system of law
18 enforcement in the examples of face recognition
19 mistakes that we have, and we have a system that
20 overwhelmingly over-investigates and over-incarcerates
21 particularly young black men. So, the system is going
22 to not operate independently of those existing biases.

1 MR. FELTEN: Thanks, Jamil Jaffer?

2 MR. JAFFER: Yeah, look, I think both Ms.
3 Bogen and Ms. Garvie have laid out a great set of
4 examples and things that we might do to address some
5 of the challenges that we face in this domain. You
6 know, I think at the end of the day, one thing that
7 ought to be considered is that, you know, these
8 systems are designed to mimic human reasoning, right?
9 They're designed to function as neural networks that
10 connect various aspects of disparate information to
11 create a holistic picture the way the human brain
12 does. So, it's not surprising that they're going to
13 have some of the similar hiccups that human brains
14 make, whether based on intentional or unconscious or
15 other forms of bias or other cognitive errors that a
16 human brain makes.

17 In some ways, the design of a neural network
18 is designed to do that. And I do worry that we have
19 in the AI domain the same kind of fetish that we have
20 in the cybersecurity domain, which is that because
21 we're talking about zeros and ones, we expect
22 perfection, right? That's not realistic in the

1 cybersecurity domain. It's not realistic in the AI
2 domain that we will not get to perfect. We will get
3 better and we're seeing that in the results over time,
4 but if we expect perfection, we'll never be satisfied.

5 And so, you're going to have to have some
6 amount of human involvement, human judgment layered on
7 top of, and admittedly, to Ms. Garvie's point,
8 admittedly deeply flawed human judgment and sometimes
9 biased human judgment layered in on top of a
10 potentially biased algorithm that's designed to mimic
11 a human brain.

12 And then the last piece of it is, the data
13 we're feeding in to train these models has its own
14 biases built in depending on how you build the data
15 and how you address the data set. So, you can bake
16 some of that out as well. You're, again, not going to
17 get to perfection, even though these are zeros and
18 ones, and these are computers, you cannot expect
19 perfection. You'll be disappointed every time, and
20 there is going to be some level (inaudible). The best
21 thing you could do is to try to train out and then
22 ultimately layer in human judgment and recognize both

1 human judgment and computer judgment are never going
2 to get you the result that you ultimately want. You
3 can only get better, not perfect.

4 MR. FELTEN: Thanks. Mr. Winn?

5 MR. WINN: Thank you for the question, Member
6 Felten. I'm as troubled as you are about why the --
7 you just continue to see abusive patterns involving
8 facial recognition, even though the algorithms are
9 getting better. I think it's because of a simple
10 mistake. And I think to some extent, Jamil pointed it
11 out and Clare and Miranda also suggested it as well.

12 People are using AI facial recognition tools
13 because they've been watching too much TV and they see
14 the word match come back. AI facial recognition tools
15 have never been intended to be used to create a match.
16 You know, for instance, if you have a chance to look
17 at the privacy impact assessment that was done for the
18 FBI's use of facial recognition technology in
19 connection with the NCIC, FBI CJIS programs, what
20 you'll see is the FBI never permits a single photo to
21 be given to an investigator. You get an array, and
22 that array doesn't go to the investigator until an

1 independent group of trained -- people who've been
2 trained in biometrics review the array.

3 The investigator is trained never to rely
4 solely on the photo or their own judgment about who in
5 the array is most likely to be the suspect, but to
6 seek corroborated evidence. And those, you know,
7 again, Dean pointed out earlier today that it's not
8 simply the machine and it's not simply the human, but
9 the rules relating to the interface between the
10 machine and the human.

11 And you've got bad rules at the state and
12 local, and it's mostly a state and local problem, bad
13 rules about how to implement and use that technology
14 against a background of a lot of biases that we humans
15 tend to have, against a framework where you're
16 thinking of this tool as being a silver bullet. Law
17 enforcement has always been about putting together
18 corroborating evidence to reduce the level of
19 uncertainty, not to achieve some perfection.

20 And I think the reason you're continuing to
21 see these instances is because people are
22 misunderstanding what the tool can do and should be

1 used for -- can be used. And they're not following
2 best practices that have been developed. And I want
3 to give the FBI a shout out on this case. They've
4 really developed some excellent tools that many
5 privacy and civil liberties organizations have
6 championed because they have shown it's the human
7 machine interface via these operational rules that are
8 constantly being evolving and improving that ensure
9 that you can use this technology in a very reliable,
10 trustworthy way where you don't destroy trust.

11 MR. FELTEN: Thanks very much. Let me pass
12 to Chair Franklin.

13 MS. FRANKLIN: Thank you. So, I want to
14 start by building a little bit on that last round of
15 questions with member Ed Felten and turn to Clare
16 Garvie. So, you have done a lot of research, I know,
17 on facial recognition, particularly in the context of
18 law enforcement. And in your opening statement, you
19 addressed some of the risks presented by the use of AI
20 for predictive purposes. And in particular, you noted
21 that many AI tools over-promise on their ability to
22 predict a propensity to commit violence or to identify

1 threatening behavior.

2 So, my question to you is, whether there are
3 any best practices that you would recommend for any
4 government agency, you know, broadening out from the
5 facial recognition context potentially, but any best
6 practices for a government agency seeking to use AI to
7 conduct any type of pattern analysis for predictive
8 purposes in the counterterrorism space? And what
9 kinds of safeguards would you want to incorporate?

10 MS. GARVIE: Thank you so much for that
11 question. I was struck by one of the questions to the
12 previous panels about what is the most serious or
13 important aspect of this vast space of AI to focus on
14 and how does the board choose that? And it did get me
15 thinking that the predictive space of algorithms does
16 seem to be one of the most critical in terms of focus
17 because of this changing who we view to be the end
18 arbiter of a decision of maybe it is anomalous
19 behavior, but maybe it's suspicious behavior, maybe it
20 is behavior that rises to the level of probable cause
21 to form an interdiction or to take negative action
22 against somebody. I do think this is where AI maybe

1 runs the risk of having the greatest harms, this sort
2 of outsourcing of guilt, if you will.

3 I think there are a couple of different
4 mechanisms to approach these tools. One is I think a
5 go, no-go analysis. Is this a place where we want to
6 automate decision-making or do the harms of that
7 decision-making or the mistakes that that algorithmic
8 system might make outweigh the benefits of moving from
9 a human or more cautious, slow approach to an AI-based
10 approach? So, that's one analysis to do before the
11 implementation of a system.

12 I think the next one is, okay, is it
13 reliable? Does it do what it says it does? And I
14 think there's far too little engagement with this
15 question before we implement advanced automated tools,
16 particularly AI in the law enforcement and other
17 spaces. We do have this inclination to see that AI
18 can solve mass data problems and then we implement it
19 without analyzing. Is it reliable?

20 So, I think that's the next check. Does it
21 reliably do what it says it does? And can we get it
22 there? Or are there -- does the human in the loop

1 solve the problem or not? I think of humans in the
2 loop as being necessary, but maybe not sufficient to
3 answer a lot of these AI questions. Just sticking a
4 human in the loop may actually exacerbate reliability
5 problems and we have to be very cautious around that
6 as a mechanism, but it is certainly a mechanism to --
7 or a lever to pull, if you will. One.

8 Another is privacy by design. What data are
9 these systems operating on? Does it have U.S.
10 person's data? I think this has to be really
11 carefully evaluated with DHS use on soft targets
12 versus maybe intelligence use abroad. DHS has now
13 partnered with Analytical AI to do anomalous event
14 detection on soft targets. I think there are very
15 real questions about whether it's appropriate for an
16 AI system to be determining what constitutes anomalous
17 or suspicious behavior at a stadium, for example. So,
18 what the target is, I think, is another mechanism,
19 another area where you have a moment to decide this
20 cost-benefit analysis.

21 And then I would look to the -- this existing
22 privacy impact assessments and systems of records

1 notices. And really urging, and this is not unique,
2 I'm not coming up with this idea, the National
3 Security Commission on the AI recommended this. AI
4 moves extremely quickly and it manifests, they changes
5 the way systems and databases operate. PIAs maybe
6 need to keep pace with that. It's not really
7 sufficient for the automated targeting system to have
8 a PIA from 2017 if it's using AI systems from, let's
9 say, 2022 and beyond. So, I think there are a number
10 of mechanisms throughout the lifecycle of developing
11 and then deploying systems that I think we need to
12 think very carefully about and pull each and every one
13 of them, depending on the harms identified.

14 MS. FRANKLIN: Thank you. Okay. I want to
15 turn next -- oh, I see fingers, I hate seeing hands.
16 Now I'm going to turn next to Miranda Bogen. So, if I
17 still have time, I'm going to give that question, and
18 then I'm going to treat those as two fingers coming
19 back to Clare Garvie. But since I'm going last in
20 this round, I'm going to turn next to Miranda Bogen
21 with a question and I'll come back to others who
22 raised their hands if I have time.

1 So, you've written about the many risks of AI
2 use and recommended that employers and others take
3 active steps to detect and mitigate or remove bias in
4 their systems. And in your opening remarks, you spoke
5 about the problems that can result from incomplete,
6 unrepresentative, or biased training data. What risk
7 mitigation tools do you recommend to avoid those harms
8 or to detect them before an AI system is implemented?
9 So, there's been talk about things like risk
10 assessments, whether those in your view have been
11 effective in testing algorithms before they're
12 deployed, and audits, which, I guess, can be either
13 before or after the fact? And what do you think is
14 the best fit for government's use of AI in the
15 counterterrorism space?

16 MS. BOGEN: Under any -- whatever label
17 people would like to call them, whether risk
18 assessments, audits, impact assessments, any proactive
19 step to review and test systems before and after
20 deployment will help to identify more issues than not
21 conducting those tests. And unfortunately, too often,
22 those tests are not conducted either before or in an

1 ongoing manner.

2 It's very important to test systems before
3 they're deployed because there could be a number of
4 design choices or even different versions of
5 algorithms that have the same quantitative results
6 that an institution might be evaluating success along,
7 but significantly different patterns of errors within
8 those results. And so, by disaggregating those
9 measurements along groups of interest, protected
10 characteristics, or other vulnerable groups, there can
11 be a comparison done to say, in pursuit of a goal that
12 we may find to be reasonable, which version of a model
13 or a system that incorporates that model best
14 accomplishes that goal while resulting in the least
15 likely harm.

16 And I think previous folks have said, you
17 know, you can't entirely remove all bias from systems,
18 there are many sources of that bias; data missingness,
19 you know, assumptions about what data is relevant,
20 that may be more pertinent for one population than for
21 another, et cetera. But by doing that type of
22 proactive measurement, you can spot whether those

1 assumptions might have led to disparities that would
2 be of concern.

3 And then continuing to conduct those measures
4 on an ongoing basis is important because, as Ms.
5 Garvie said, the conditions of deployment may
6 significantly differ than the conditions of testing.
7 And so, unless you're testing that in the wild and
8 also understanding how human are acting on the output
9 of systems to the extent that they're being relied
10 upon to do so. We won't know if -- even if there
11 might be no disparities in the performance of the
12 system, which is highly unlikely, but even if there
13 were, whether the humans are acting differently in
14 similar circumstances in a way that would need to be
15 identified and for which processes would need to be
16 implemented to prevent that from happening.

17 So, again, very simple approaches, but alas,
18 don't tend to be prioritized across the board. And
19 so, whether they're incorporated into impact
20 assessment, risk assessment, audits, risk management
21 processes, those types of considerations are necessary
22 alongside considerations of overall accuracy of a

1 system independently.

2 MS. FRANKLIN: Thanks. Okay. I think I have
3 2 minutes left before our final lightning round among
4 back to the board members. So, I just want to give a
5 chance to Peter Winn and Jamil Jaffer, you know, super
6 quick, what you raised your hands for just before?

7 MR. WINN: Just the point of the -- doing
8 privacy impact assessments after the system and the
9 operational procedures have been established is
10 useless. You have to do them early -- during the
11 early development phase of the process. And then,
12 what inevitably happens is, you know, you're trying to
13 predict as much stuff as you can, you can't predict
14 everything. I look back on privacy impact assessments
15 I signed in 2017 and I'm appalled today at all the
16 things I've missed. So, you have -- it's a continuous
17 process, you have to have those privacy impact
18 assessments going back and looking at what you now
19 know and, you know, and then you do it again because
20 otherwise, you're really going to lose the benefit of
21 all the knowledge that you're gaining through, you
22 know, your ignorance. You know, you're mitigating

1 your ignorance and you're not -- you have to go back
2 and cycle this risk mitigation process. It's a
3 continuous --

4 MS. FRANKLIN: Thank you. I'm just giving 30
5 seconds to Jamil Jaffer for what he was raising his
6 hand for.

7 MR. JAFFER: You know, look, I think Ms.
8 Bogen, Ms. Garvie, and Mr. Winn said almost exactly
9 the same thing, which is that you got to do this from
10 the beginning, bake it in from the beginning and then
11 to do all the way through. But the key is, if you're
12 just the government doing and saying, we're going to
13 do reassessments, it's not going to work, you've got
14 to incentivize industry players and investors who are
15 already incentivized in their own ways to do these
16 things. And the core there is this idea that trust,
17 safety, and security actually benefits your return on
18 investment, it benefits the intellectual property that
19 you're creating, it benefits the uptake of these
20 capabilities. And the more the government can meet
21 that part of their buying mechanism, and part of their
22 feedback to industry and investors, that's really

1 what's going to drive this thing. It's not going to
2 happen because you do a bunch of PIAs over in the
3 government. The key is to bake this in and have a
4 continuous assessment process go on.

5 And heavy-handed regulation by the way, is
6 simply going to suppress innovation. What you really
7 want is incentivization of the right kind, but also
8 gives you the answer that we all want, which I think
9 everybody agrees on.

10 MS. FRANKLIN: Okay. Thanks. Okay. So,
11 final lightning round back up to Board Member Beth
12 Williams.

13 MS. WILLIAMS: Great. Thank you. So, one of
14 our former board members, Jim Dempsey, has written
15 extensively on the importance of contestability in AI
16 systems. And so, I'm wondering if in this lightning
17 round, you can just quickly tell us if you think
18 contestability is an important consideration and how
19 you think we can best incorporate contestability into
20 some of these systems. And so, I think because I'm a
21 Williams, I'm going to ask to go in reverse
22 alphabetical order, starting with Peter Winn.

1 MR. WINN: Well, thanks. Jim Dempsey is one
2 of my favorite PCLOB board members and currently one
3 of my favorite privacy, data protection review court
4 judges. So, he always has very thoughtful things to
5 say. I think that contestability ideas that he has
6 also involve asking the right questions and probing in
7 the right ways, and a multi-stakeholder process. And
8 that's also described in the intelligence community's
9 framework for IC development.

10 The importance of having multi-stakeholders
11 engaged in pushing and asking those questions from
12 lots of different perspectives because none of us have
13 that monopoly or knowledge that we all wish we had.
14 And bringing in that multi-stakeholder process to the
15 extent you can and you can -- even in a classified
16 environment, you can bring in a lot of multi-
17 stakeholders. The PCLOB itself represents a very
18 diverse body of board members representing, you know,
19 a similar kind of diversity of views and judgment,
20 pressing all of the aspects of the development of
21 these programs. Looking at the underlying data,
22 understanding how the algorithms work, all of that is

1 critical. But it can't be done by a single point of
2 view. It has to be done in a multi-stakeholder way.

3 MS. WILLIAMS: Thank you. Mr. Jaffer?

4 MR. JAFFER: Yeah, look, I think, obviously
5 contestability and being able to push back against a
6 decision made by AI for an individual is the right
7 thing to do. I can't imagine anybody on this panel is
8 going to disagree that you should have contestability
9 baked in. And so, to me, you know, the Dempsey,
10 Landau idea of contestability is exactly the right
11 one. I do want to say something to the data
12 protection court, which is that (inaudible)
13 contestability in America -- in the American system,
14 it should be for Americans. This idea that we're
15 bringing Europeans in, and we're giving them this fake
16 court made up of executive branch officials is
17 ridiculous, and completely antithetical to our system
18 and makes no sense whatsoever. So, I did want to put
19 that out there. I do love Jim Dempsey, Data
20 Protection Court, ridiculous.

21 MS. WILLIAMS: Thank you. Ms. Garvie?

22 MS. GARVIE: Yes, just echoing

1 contestability, super important on -- in at least two
2 aspects. One is the foundational validity. Does this
3 work as intended? Have we tested it? And have we
4 tested it sufficiently and independently? And then
5 the validity as applied aspect as well. If it didn't
6 go right or wrong in this particular case, and can the
7 person directly affected, challenge that and contest
8 that?

9 MS. WILLIAMS: Thanks. Ms. Bogen?

10 MS. BOGEN: Yes, I was honored to participate
11 in the series of workshops that led to the report on
12 contestability by Mr. Dempsey and Ms. Landau. And so,
13 I would certainly endorse the findings of that
14 workshop. And I think one of the main conclusions
15 that came out of that is contestability is not an
16 independent concept from due process. And so, we need
17 to remember all of the circumstances under which due
18 processes is guaranteed, and the introduction of AI
19 does not change that. I think we also need to be
20 attentive to the limitations of these systems and the
21 justifications they can or can't make around the
22 recommendations and ensure that human analysts, again,

1 similar to the case of facial recognition, don't
2 solely rely on the output of AI systems to justify
3 actions that would otherwise not be justified without
4 corroborating evidence or signals.

5 MS. WILLIAMS: Now, turn it over to Member
6 LeBlanc.

7 MR. LeBLANC: Thank you, Beth. Mr. Winn, I
8 hesitate to suggest that I might also have read some
9 of your early PIAs and wondered a few times what you
10 were thinking at that time. And I won't ask you to
11 tell us which ones of those are outdated so that you
12 can correct them. But I do want to go back to the
13 question that I had posed to Mr. Jaffer in the last
14 round, because I did notice that you came off mute
15 right after he finished his comments. And it was
16 about the FBI's prior errors in the FISA context. And
17 of course, recognizing that the bureau is a component
18 of the Department of Justice, it's only fair to give
19 you an opportunity to share any remarks about how the
20 FBI can build trust in its use of AI.

21 MR. WINN: Thanks -- thank you, Member. I'm
22 very grateful for the opportunity to respond. And

1 we'll have a separate conversation about which ones --
2 which PIAs that I signed in the past are most in need
3 of updating. The -- but, you know, we've been talking
4 about the machine, we've been talking about the human,
5 and we've been talking probably not enough about the
6 interface. And the interface being the rules that
7 apply when you're deploying the human and the machine
8 to accomplish a mission. And your question earlier
9 about 12333, the guidelines, the attorney general
10 guidelines that are issued pursuant to 12333 and at
11 the FBI, I would highlight how they in turn implement
12 the attorney general guidelines through the DIOG or
13 the Domestic Investigations Operation Guide. That's a
14 massive standards for good law enforcement, okay?

15 I would say that, you know, Jamil was talking
16 about the difference between accidents and on purpose.
17 And even a dog knows the difference between being
18 tripped over and kicked. When the FBI -- and the
19 Durham report which was issued by Special Counsel
20 Durham, discussing a breakdown in trust at the FBI.
21 If you read it carefully, you can see that what he
22 points out is the intentional violation of their own

1 rules, the DIOG. Now, the DIOG isn't required by
2 statute, isn't a regulation, but it is an -- for
3 years, it was the FBI's Bible. It was what made you
4 an FBI agent and why the FBI was so much better than
5 any other law enforcement agency at what it's -- it
6 was a self -- it was the identity of what it meant to
7 be an FBI agent. And the DIOG was established by --
8 originally by Attorney General Edward Levi to deal
9 with the breakdown of trust that took place under
10 Edward -- you know, Director Hoover with the
11 COINTELPRO scandals where they were going up on Dr.
12 Martin Luther King, you know, based on not enough
13 evidence.

14 And the DIOG was put in place with thresholds
15 of evidence, you can't open an investigation with just
16 a little evidence. You can do an assessment, you have
17 to do a preliminary investigation and only when you
18 get more evidence, then can you open up full
19 investigation. And only when you have a full
20 investigation, can you actually get a wiretap or
21 something like that. That's -- those things are baked
22 in. They are the interface. They are the rules that

1 don't go away when we bring in AI. And the mistake
2 that people are making is they're thinking those rules
3 don't apply anymore, then we do our --

4 MS. FRANKLIN: Thank you. Peter? I'm so --

5 MR. WINN: -- we're checking to make sure
6 those rules are still being followed, and trust is
7 maintained. And so, that's come to a conclusion.

8 MS. FRANKLIN: Thank you. Sorry, just trying
9 to make sure that Ed Felten and I get a chance for our
10 last lightning round question. So, over to Ed Felten.

11 MR. FELTEN: Okay. Yes, I want to come back
12 to a question that Mr. LeBlanc asked Ms. Bogen
13 earlier. And that is about how the general frameworks
14 for AI governance, such as the NIST framework might
15 need to be adjusted or augmented in the context of
16 counterterrorism. And in the interest of lightning
17 round efficiency, I will ask each of the other three
18 panelists who have not yet addressed the question to
19 give a brief answer on that topic. And I'll start
20 with Mr. Winn.

21 MR. WINN: So, the brief answer is we've got
22 to focus not simply on doing the general risk

1 assessments that NIST is talking about. But we have
2 to make them context specific. And we have to look at
3 our prior rules of engagement, the rules that have
4 been developed through knowledge and experience of
5 generations of law enforcement and national security
6 individuals. Those rules need to be baked in as well
7 to the risk assessment process.

8 MR. FELTEN: Thanks. Mr. Jaffer?

9 MS. JAFFER: Yeah, I mean, look, the AI risk
10 assessment NIST, you know, frameworks are frameworks.
11 They're not designed to be the exact thing you
12 implement every day, day to day in and out. They're
13 designed to be customizable to a variety of contexts.
14 And so, I think in the government context, you ought
15 to apply them in a way that makes sense. And that
16 accounts for the unique issues that Peter and Ms.
17 Bogen and Ms. -- and the other panelists have -- and
18 Ms. Garvie have raised as well. So, I think just --
19 you got to apply the frameworks in a contextual way.
20 So, I don't think there's anything surprising there.

21 MR. FELTEN: Right. Ms. Garvie?

22 MS. GARVIE: Agree. And I would also maybe,

1 this is pie in the sky, but I would love to see the
2 intelligence and national security community also
3 adopt something that DARPA is now adopting from the
4 genomics project, which is broadening it a little bit,
5 the risk framework a little bit out to ethical,
6 social, and legal implications. So, a little bit
7 broader than just privacy because I think that helps
8 anticipate potential future problems or challenges and
9 concerns caused by AI systems to the point that we've
10 been talking about earlier with these PIAs getting out
11 of date so quickly. And just as an example of how we
12 need to anticipate from a broader perspective, the
13 implications of these systems.

14 MR. FELTEN: Great. Thank you to all the
15 panelists. And I'll pass to Chair Franklin.

16 MS. FRANKLIN: Thank you. Okay. So, final
17 question to close us out. If you can, each, I'm going
18 to go through in forward alphabetical order again. If
19 you can each give us a concise framing as you can of
20 how would you scope and define an appropriate slice or
21 focus for PCLOB's oversight of AI in counterterrorism?
22 So, starting with Miranda Bogen.

1 MS. BOGEN: I'm sure the other panelists will
2 have very insightful perspectives on the question
3 itself. So, I will just say no matter what slice
4 PCLOB chooses to focus on, you should also make
5 recommendations about to the extent there are other
6 elements that PCLOB is not going to focus on or it
7 isn't within their ambit to, that other analogous
8 oversight mechanisms are set up to focus on those
9 other elements.

10 MS. FRANKLIN: Thank you. Clare Garvie?

11 MR. GARVIE: I think it's less a question of
12 what the appropriate slices and more sort of a
13 hierarchy. But my current hierarchy in the course of
14 this conversation, I think would place predictive
15 systems where we've supplanted human decision making
16 with an automated decision sort of at the top of that
17 hierarchy, followed probably by systems where AI is
18 being implemented into an existing structures,
19 particularly mass datasets, and fundamentally changing
20 the nature of the data and its applications where the
21 impact assessment has been already assessed at
22 collection, but AI is changing the applicability,

1 usability of that data into something that's new and
2 raises new challenges.

3 MS. FRANKLIN: Thank you. Jamil Jaffer?

4 MR. JAFFER: Yeah, I think the key focuses
5 for PCLOB has to be staying within a statutory
6 mandate, right, which is about efforts to protect the
7 nation against the threat of terrorism, right? What
8 you don't want to do is end up with the PCLOB on a
9 roving search for AI challenges with national security
10 more generally, right? What if the PCLOB's mandate
11 was broader, right, that's a different question for
12 Congress to consider if they want to broaden your
13 statute. But to the extent that they've given you a
14 statute you have, you've got to stay within the
15 counterterrorism construct. And so, to the extent
16 that AI is being used in the counterterrorism mission
17 space, or is going to be used in that space, that's a
18 place for PCLOB to focus. I don't think there's
19 necessarily a specific slice within that. But staying
20 focused on the counterterrorism mission, and not
21 getting into the related national security matters, I
22 think will be the thing that allows PCLOB to do its

1 job most effectively.

2 MS. FRANKLIN: Before we go to Peter Winn, I
3 just have to say, our jurisdiction clearly covers
4 multiple purpose programs and activities that include
5 counterterrorism with Section 702 being --

6 MR. JAFFER: We can debate that. We can
7 debate that. I'm not sure that's the right way of the
8 statute.

9 MS. FRANKLIN: Over to Peter Winn.

10 MR. WINN: I'm not going to get into the
11 jurisdictional debate. But I know that as Member
12 LeBlanc said, domestic terrorism is a serious concern.
13 And the domestic terrorism context in the United
14 States is done through the law enforcement structures.
15 The law enforcement structures have well developed
16 rules, like the FBI DIOG. I would look at this
17 question, which is, is the law on the books, law on
18 the ground? Are people actually doing what they say
19 they do? And if they're not, you need to hold us
20 accountable. And the PCLOB is in a special position
21 to make transparent, both the things that we're doing
22 right as well as the things that we're doing wrong, so

1 the public better understands how to evaluate whether
2 this is done on purpose or whether it was an accident.

3 MS. FRANKLIN: Okay. Thank you all and
4 thanks to all our panelists for sharing your insights
5 with us today. And thank you to everybody who has
6 been joining in our audience and this will close us
7 out. Thank you.