



ADMINISTRATIVE CONFERENCE OF THE UNITED STATES

ARTIFICIAL INTELLIGENCE IN THE TRENCHES
A View From The Agencies

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TRANSCRIPT
(Not Reviewed for Errors)

Panelists

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Kurt Glaze, Program Analyst, Office of Analytics, Review, and Oversight, Social Security Administration

Sean Khozin, Global Head of Data Strategy, Janssen R&D; Formerly Associate Director, Oncology Center of Excellence, Food and Drug Administration

Krista Kinnard, Director, AI Center of Excellence, General Services Administration

Moderator

Stephen Sanford, Director, Center for Strategic Foresight, Government Accountability Office

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Transcription of Audio File:

Artificial Intelligence In The Trenches_ A View From
Inside The Agencies

Audio Runtime: 59:04

1 (Beginning of audio recording.)

2 MR. WIENER: Good afternoon. I'm Matt Weiner,
3 the vice chair and executive director of the
4 Administrative Conference of the United States, ACUS
5 for short. Welcome to the fourth and final panel of
6 our symposium on AI in federal agencies, which ACUS
7 has been pleased to cosponsor with Georgetown's
8 Institute for Technology, Law, and Policy.

9 Today's panel, like the prior three panels,
10 today's panel will be both recorded and transcribed.
11 The video and transcription will soon be available on
12 the ACUS website.

13 In the first three panels, we largely heard the
14 perspectives of experts outside federal agency, eight
15 agencies today. We'll hear from several innovators in
16 and practitioners of AI within federal -- inside
17 federal agencies, and we're very thankful that Stephen
18 Stanford, director of the Center for Strategic
19 Foresight at the Government Accountability Office,
20 GAO, has joined us to moderate our discussion today.

21 I'll turn things over to the Institute's Jeff
22 Gary in a minute. I'd first like to thank, though,
23 him and the Institute's director, Hillary Brill
24 (phonetic) for cosponsoring this symposium with ACUS,
25 ACUS attorneys Todd Phillips and Todd Rubin for

1 organizing this symposium so expertly, all of our
2 panelists, of course, and especially our many
3 attendees, 200 today I'm told, many themselves experts
4 in AI.

5 With that, let me turn it over to you, Jeff, with
6 my thanks once again.

7 MR. GARY: Absolutely. Thank you, Matt, and I'd
8 also just like to extend a great thanks to ACUS and
9 especially to Todd Phillips and Todd Rubin, who have
10 been doing just phenomenal work on this, planning even
11 before we switched to a webinar format.

12 Everyone, I'm Jeff Gary. I'm a project manager
13 at the Institute for Tech, Law, and Policy at
14 Georgetown. The Institute is a think tank. It's run
15 through the law center, and we focus deeply on some of
16 the cutting-edge questions raised by new and emerging
17 technologies. And we've been so pleased that we can
18 be a part of bringing this program to life with ACUS.

19 You know, we believe strongly that as AI and new
20 technologies develop, they bring new opportunities,
21 and we've been able to explore those. But at the same
22 time, we can't dismiss the challenges that these
23 technologies pose to existing social issues such as
24 systemic racism and discrimination and the lack of
25 social mobility.

1 And I really feel that this series has shed a
2 strong long on really the nuance and the realities of
3 utilizing and engaging with these technologies, and
4 we've had a really strong and open discussion about
5 where things are working and where things are not and
6 the benefits and challenges faced by agencies.

7 And with that in mind, I'm so glad that our panel
8 today focuses on the realities of the agency's use of
9 these technologies and how they're really grappling
10 with them.

11 We'd love to stay in touch. So please do follow
12 us on Twitter. We're at Georgetown Tech Law -- sorry,
13 Gtown Tech Law. And visit our website, which is
14 GeorgetownTech.org, where you can sign up for a
15 newsletter if you want to attend other events like
16 this.

17 On that note, I'm going to turn things over to
18 Steve Sanford, who's going to introduce our panelists
19 and get things moving into the discussion. Thanks
20 again, everyone.

21 MR. SANFORD: Thank you very much, Matt. Thank
22 you very much, Jeff, for those intro remarks. Very
23 pleased to be here today. My name is Steve Sanford.
24 I am the director of the Center for Strategic
25 Foresight at the U.S. Government Accountability

1 Office. Much like ACUS, we are interested in
2 improving the efficiency and the effectiveness of the
3 federal government. So that's certainly a mission
4 that, as an independent, nonpartisan agency of the
5 legislative branch, we feel very akin to.

6 Very pleased to be joined by the panel today. In
7 my work at GAO, I led GAO's first technology
8 assessment on AI in 2018. I've done some work on deep
9 fakes, and I'm currently working on a program of work
10 related to AI governance, and the Comptroller General
11 is going to convene an expert forum this September and
12 will be subsequently issuing some findings from that
13 work. So we're really excited to be a part of this
14 panel today.

15 And without further ado, I would like to
16 introduce our panelists. So I'm going to ask our
17 panelists just to say a few words about themselves.
18 Let's start with Krista, please.

19 MS. KINNARD: Yes. Hello, everyone. Thank you
20 so much for having me. It's really a pleasure to be
21 here. I'm the director at the Artificial Intelligence
22 Center of Excellence, housed within the Technology
23 Transformation Services at GSA. And so because we sit
24 in GSA, we really -- we have a government-wide focus.
25 And our main mission is to partner with the rest of

1 the federal government and across federal agencies to
2 help accelerate the adoption of artificial
3 intelligence in government agencies. And we do that
4 in a number of ways, both through the sharing of best
5 practices and lessons learned in our federal AI
6 community of practice and also through direct
7 partnership through our centers of excellence program.
8 So thank you very much.

9 MR. SANFORD: Thank you, Krista. Welcome. Sean,
10 please.

11 MR. KHOZIN: Hi, everyone. I'm Sean Khozin. I
12 must highlight the fact I've recently left the federal
13 government. So I believe I'm the only one here who is
14 not currently in federal government.

15 So currently, I'm the global head of beta
16 strategic, data science innovation at Johnson &
17 Johnson Pharmaceuticals, and a few months ago before
18 joining J&J, I was at the FDA as associate director of
19 the (inaudible) Center of Excellence and also as a
20 founding director of Information Exchange and Data
21 Transformation that we launched in 2015 with special
22 authority from the Department of Health and Human
23 Services as an incubator at the FDA for developing new
24 capabilities in artificial intelligence and other data
25 science-related areas.

1 I'm a thoracic oncologist, and I've been doing
2 data science academically and also in the private
3 sector for a number of years going back to my
4 (inaudible). Proud to be part of this conversation.

5 MR. SANFORD: Thank you, Sean. Kurt, please.

6 MR. GLAZE: Yes, thank you, Stephen and ACUS and
7 the Institute for having this panel today, and I'm
8 honored to be here. My name is Kurt Glaze. I
9 formerly was an appellate attorney at the Social
10 Security Administration, and now I am a program
11 analyst there with the Analytic Center of Excellence
12 within SSA. I'm also the creator and lead of a
13 decision support application project known as Insight,
14 which is a software product designed to help
15 disability program adjudicators with their
16 adjudication activities at the hearings and appeals
17 levels of adjudication. So thanks again for having
18 me.

19 MR. SANFORD: And finally Marco, please.

20 MR. ENRIQUEZ: Hey. My name is Marco Enriquez.
21 I'm a senior applied mathematician in the U.S.
22 Securities and Exchange Commission. I run the largest
23 analytics program at the SEC that's (inaudible). We
24 utilize and evaluate AI technologies in support of the
25 SEC's mission.

1 Also, I'd like to add thank you to ACUS and also
2 the views that I'll express today are my own and don't
3 reflect the commission's. Thank you.

4 MR. SANFORD: Thank you, Marco. Appreciate that.
5 I'll add my own disclaimer that any remarks that I
6 make today are also my own and don't necessarily
7 reflect the institutional views of GAO. So thank you,
8 Marco, for that reminder.

9 Let's hear next from all of our panelists. I'd
10 like to hear from them each in series. And if you
11 could tell us a little bit about your organization's
12 journey and AI and machine learning -- where you
13 started and where you are now, what specific ways your
14 agency is using AI and machine learning to fulfill its
15 mission. And as you've been going through that, what
16 have been some of the key enablers of success as that
17 journey has been underway?

18 So again, we'll start with Krista, please, if you
19 could do that. And I'd also invite our audience,
20 who's listening live, please feel free to post
21 questions during these initial comments from the
22 panelists. We'll do a quick follow-up round after
23 these remarks to answer some of those questions, and
24 then we'll do more questions at the end. But we do
25 want this to be interactive. So if you have any

1 questions as you're hearing from our panelists, please
2 put those forward. So we'll be hearing from each
3 panelist for about five minutes. So starting with
4 Krista, thanks.

5 MS. KINNARD: Great. Thanks so much, Stephen.
6 So the Artificial Intelligence Center of Excellence
7 and the broader (inaudible) AI portfolio is a very new
8 organization. It was set up last October. So we're
9 almost a year old in light of the executive order on
10 maintaining American leadership on artificial
11 intelligence.

12 And so we sit in a part of GSA called Technology
13 Transformation Services, whose entire mission is to
14 engage with the rest of the federal government, as I
15 kind of alluded to earlier.

16 So we're not really doing development within
17 GSA's internal programming. We are a program that
18 exists to partner with the rest of the federal
19 government. And so we do that in a number of
20 different ways.

21 And so really the whole genesis of this and why
22 we have an Artificial Intelligence Center of
23 Excellence was that agencies are excited -- and I
24 think we're about to hear from a group of folks who
25 are going to show you that agencies are already

1 starting to embrace artificial intelligence. And I
2 think that there's a lot of really interesting and
3 engaging use cases around how this technology can
4 really be applied to help agencies meet their mission.

5 But there's a lot of questions about it, right?
6 I think everyone here would agree that AI is not a
7 solved technology, right? You can't just buy one AI
8 and get it out of the box. It's complicated, it's
9 nuanced, and as agencies start thinking about, you
10 know, what are my principal business and mission
11 challenges that I'm trying to meet, how can AI as a
12 technology be used, and if I really want to use
13 artificial intelligence to support my mission, what
14 does that look like? What do I do? Right? And so
15 that's why we exist.

16 And part of what we do, right, so part of what we
17 established when we set up this group is a community
18 practice. Because first and foremost, what we know is
19 that we're not in this alone, right? And going off
20 and trying to build individual artificial intelligence
21 systems in a black box is not going to serve anyone.

22 So we wanted to create a space for the federal
23 government to come together. And we do through a
24 series of webinars. So far, we've led a series on
25 acquisition in AI, right? I mean, there are some

1 agencies who are going to develop a lot of this in-
2 house. They've got deep technical staff to be able to
3 do this.

4 But if we're speaking realistically, acquisition
5 and procurement is going to be at least part of an AI
6 solution. So really trying to figure out how can the
7 federal government be smart about what we're buying,
8 right, is really something that our community wanted
9 to learn about.

10 The other thing that we've done, we've done some
11 tech talks, right? We had a Ph.D. computer scientist
12 come in and start talking about how you can start
13 using some open sourced tools to start playing around
14 with data in your organization.

15 And really the goal here is to bring people
16 together. And I have just been overwhelmed by the
17 number of people who are just really excited, and
18 they'll say something like, oh, I heard such and such
19 agency can speak to your community. I'm trying to do
20 something similar. Can you connect me, right, so that
21 we can -- we can share those lessons learned and share
22 -- there are other things, right? Share resources,
23 share tools, share frameworks, right?

24 And our community is incredibly strong. We have
25 almost 1,000 members, and we are less than a year old.

1 And they're really engaged, which I think is -- is
2 incredibly powerful but also speaks to the demand and
3 the interest in bringing this technology to the table.

4 And to be honest with you, I've just been
5 incredibly impressed by partner groups who have spoken
6 up, and they just -- they ask such good questions
7 about things like how do you do this responsibly, how
8 do we mitigate bias, right, how do we start thinking
9 about privacy when we are building these solutions,
10 right?

11 And I don't know the answer to all of those,
12 right. And really the community isn't here to solve
13 your problems. The community is here to bring those
14 to light so we can solve them and start addressing
15 them together.

16 The other side of what we do through the centers
17 of excellence is we do actually partner, right, one-
18 on-one with agencies, and more broadly our centers of
19 excellence program focuses in six different areas. So
20 we have our Artificial Intelligence Center of
21 Excellence, which is our newest, but we also have a
22 longstanding Data and Analytics Center of Excellence
23 infrastructure, cloud, customer experience, and call
24 center modernization, right?

25 And we have kind of this demonstrated history of

1 being able to partner one-on-one, bring in our
2 technical expertise to drive projects. And we engage
3 on an executive level because we know that for these
4 types of projects to really gain traction, to really
5 have the impact that I think that they really should
6 and can have in a way that is meaningful, in a way
7 that is bringing value, in a way that is responsible,
8 you need to have that championship, right? You have
9 to have those leaders stepping up and saying this is
10 something that is important to our organization, and
11 this is something we want to do, and it's something we
12 want to do well.

13 And that's really where our centers of excellence
14 thrive, begin able to get in there, roll up our
15 sleeves, and with an organization, speak to their
16 executive leadership and go across the organization to
17 say we need these partners to come together to
18 actually do an implementation.

19 So I lead one of our engagements with the
20 department, and we're doing some automation and
21 bringing in intelligence to some of the processes that
22 they're looking at kind of in their procurement cycle.

23 So I'll stop there. I know I've taken up a lot
24 of time, but I -- I'm happy to field questions about
25 what we're seeing in the federal landscape, what are

1 some of the common challenges not just in one specific
2 agency but across agencies, and where are agencies
3 both finding opportunities and challenges to come
4 together and really make an impact in artificial
5 intelligence adoption.

6 MR. SANFORD: Great, thank you, Krista. It's
7 remarkable to hear about some of the momentum right
8 now behind AI and the awareness of the issues in the
9 federal space, maybe even just compared to two or
10 three years ago. It sounds like there's been quite a
11 lot of movement there, and I'm sure we'll come back to
12 some of those issues you've raised.

13 Next, let's hear from Sean, who's seen both the
14 experience of how machine learning is deployed on the
15 federal side with FDA and then also in the private
16 sector. So Sean.

17 MR. KHOZIN: Sure. So the FDA, when we look at
18 the application of AI and machine learning at the
19 agency, I think one can divide it into three different
20 categories. The efforts that are aimed at automating
21 workflows and business process. So there are a few
22 pilots that, when I was at the FDA, I was involved in
23 and that are being entertained right now.

24 And those are some of the same themes that any
25 organization and any business can take advantage of.

1 And in fact, at J&J, we're also applying machine
2 learning and piloting some of these methods in
3 streamlining and potentially automating certain
4 business processes. So that's one category.

5 And the second category is the way that one
6 applies machine learning and AI to analyzing
7 biomedical data assets. For the FDA, that's about,
8 you know, approving drugs (inaudible) devices. And
9 for us at J&J, it's about extracting insights from the
10 data we're generating as part of our trials and
11 developing programs.

12 What's interesting is that, you know, the FDA in
13 a lot of cases have been applying modeling and
14 simulation and more recently methods that can
15 categorize as AI and machine learning in very specific
16 contexts.

17 A great example is pharmacometrics.
18 Pharmacometric review of drugs involves a lot of
19 modeling and simulation, and the FDA does have great
20 expertise in that area, and they've been increasingly
21 applying machine learning methods.

22 And again, we are also applying machine learning
23 to a lot of the pre-clinical discovery work that's
24 being done. And (inaudible) component, and there's
25 been several publications. One was by the FDA, and I

1 led several AI efforts on the academic front that
2 produced published academic papers that we felt could
3 advance the priorities of the FDA, the mission of the
4 FDA, and also provide actually insights to the
5 industry in terms of, you know, the art of the
6 possible, in some cases.

7 Because FDA is in a very unique position where it
8 has access to essentially all the drugs and biologics
9 that have been ever approved in the United States, and
10 that data can be leveraged in very unique ways and a
11 perfect substrate for AI.

12 And then the third component of that is how the
13 FDA approaches the approval of AI algorithms, and
14 there are several pilots there. For example,
15 (inaudible) has (inaudible) that address that, and
16 there are a number of different programs that are
17 underway. And as many of you may have heard,
18 (inaudible) technology modernization plan that is
19 aimed at providing a framework for scaling data
20 science and also AI machine learning across the
21 agency.

22 For Jansen and J&J as healthcare and
23 pharmaceutical company, we are incorporating
24 (inaudible) analytics, including machine learning
25 methods, across essentially the entire drug

1 development continuum and bringing in new talent and
2 building infrastructure, and that was one of the many
3 reasons that I joined the company and was really
4 providing valuable insights to (inaudible) some of our
5 programs and also providing an opportunity to just ask
6 fundamentally different questions that one typically
7 doesn't ask as part of, you know, traditional drug
8 development paradigm.

9 And these methods do allow researchers and drug
10 developers to really ask fundamentally different
11 questions that without applying such methods wouldn't
12 -- we wouldn't be able to answer.

13 MALE VOICE: But once you justify handing them
14 out to law students, you can't justify not handing
15 them out to anyone basically who wants one.

16 MR. KHOZIN: Yes, was that a question? I believe
17 I heard a question.

18 MR. SANFORD: I'm not sure if that was -- I'm not
19 sure who asked the question. Well, so Sean, was that
20 the end of your remarks, Sean?

21 MR. KHOZIN: Yes.

22 MR. SANFORD: Great, thank you. Thank you very
23 much. So Kurt, next from you, you already mentioned a
24 little bit that you're actively deploying some tools
25 in the fulfillment of the mission. So I would love to

1 hear about that.

2 MR. GLAZE: Sure. So I suppose before I begin,
3 I'll give a little bit of back story. I started my
4 career as an appellate attorney at SSA for its
5 disability program. And then in 2015, after years of
6 engaging with that program from, you know, a purely
7 legal side, I pitched software -- a concept for
8 software, and that's now called Insight, and that is
9 my current major project, and I've been working on it
10 in more or less fulltime since 2015.

11 So some background on Insight is essentially in a
12 nutshell, it is decision support software, again,
13 designed to provide adjudicators with a series of
14 things. First of all, it reads the text of disability
15 decisions written at the hearing, written or reviewed
16 at the hearing or appeals levels of the disability
17 program. We have multiple levels of adjudicative
18 review.

19 It extracts information from the (inaudible)
20 texts of those decisions, combines it with other
21 information about the case and claim in our current
22 systems, and ultimately offers feedback to
23 adjudicators about potential quality issues that are
24 present on the face of the decision.

25 In addition to that, we offer them contextual

1 reference information. So you know, kind of targeted
2 reference information that is specific and helpful for
3 that claim only. And we also offer them a series of
4 tools that are basically enabled through the
5 information that we extracted upstream.

6 All of this is designed to help adjudicators
7 improve the quality, efficiency, and consistency of
8 our disability decisions in service to the public.

9 So I think to understand the why as to why
10 Insight, you know, we even rolled this out or why SSA
11 funded the development of Insight, it helps to kind of
12 understand how SSA adjudicated claims prior to
13 Insight, and I think the SSA story pre-Insight is very
14 common among mass scale adjudicative agencies in the
15 federal government.

16 So prior to Insight, the workflows essentially
17 consisted of individual attorneys at either the
18 hearings or appeals levels working individually on a
19 case, you know, preparing a work product, and kind of
20 proactively seeking out, based on their training and
21 experience, the resources they need, looking up the
22 right regulatory or subregulatory guidance for the
23 claim, and then passing their completed work product
24 to another lawyer or adjudicator for independent
25 review by them.

1 And if a quality issue was found or additional
2 work needed to occur, that other individual would send
3 it back to the first individual. But -- so you guys
4 can glean that this was a very manual process,
5 independent action by independent action.

6 So how Insight breaks that paradigm is it
7 intervenes at the individual level to try and
8 proactively bring relevant information to the
9 individual rather than them constantly having to seek
10 out that information.

11 It also, as I said, its probably most notable
12 feature is its capacity to potentially flag quality
13 issues that merit further attention by that individual
14 before they push their work product down the line.
15 And as you all can imagine, it is much more
16 inefficient to deal with these quality issues if
17 they're embedded in work products downstream such as
18 on appeal, or after the work product has been handed
19 off to another staff person than it is to address them
20 as soon as possible when the worker is right there
21 looking at, you know, potentially thousands of pages
22 of evidence in an individual case and trying to digest
23 all of this.

24 So that is the aim of Insight. As you mentioned,
25 Insight is fairly far along. It's not conceptual.

1 We've actually deployed Insight software to all
2 adjudicators and adjudicative staff, by which I mean
3 essentially kind of like law clerks, attorneys who
4 write decisions, or at the appeals level, appellate
5 attorneys who conduct an initial review of an appealed
6 case.

7 We've rolled that software out for use by all
8 attorneys, essentially, at the hearings and appeals
9 level, and that's been the case since essentially --
10 at both those levels since 2018.

11 I'm happy to discuss this later, but we've also
12 conducted several analyses of Insight's impact in
13 business value for Insight's disability program and
14 have seen some positive results both in improvements
15 to the quality of the decisions as far as we can tell
16 but also to the efficiency of adjudication, which at
17 SSA I can tell you as such a large body is extremely
18 important that we do whatever we can to serve the
19 public efficiently.

20 I think some of the most important success
21 enablers that led to Insight -- Insight wasn't
22 developed in a vacuum. We benefitted at SSA from
23 having an electronic system, which some adjudicative
24 bodies are, you know, more advanced or less advanced
25 stages.

1 We also had a fairly robust case -- electronic
2 case processing system, and both of these systems
3 upstream were kind of the foundational building blocks
4 upon which Insight could possibly build some of the
5 technologies that it has.

6 SSA is also -- in this recent decade underwent
7 many efforts to globalize data and enterprise data
8 warehouses and otherwise make experimentation more
9 feasible for teams across the agencies.

10 So some of those have been major enablers for us
11 that, you know, we stood on the shoulders of those to
12 build Insight. So once again, thank you for having
13 me.

14 MR. SANFORD: Thank you, Kurt. I appreciate
15 those remarks. Sounds like you've come a long way in
16 that journey from 2015 and continuing to move forward.

17 Marco, please let's hear from you.

18 MR. ENRIQUEZ: Thank you. So like Hurt, allow me
19 to just maybe give a little bit of background. So my
20 interest with machine learning really just started in
21 graduate school. I have a doctorate in applied
22 mathematics, and the dissertation topic that I chose
23 was in the field of optimal control theory, which is
24 really the precursor to what people call reinforcement
25 learning in the AI community.

1 And when I graduated, I joined industry around
2 the time where big data was the latest buzzword, and
3 it was really an exciting time because everybody was
4 trying to really analyze and parse like hundreds of
5 gigabytes and terabytes of data and also considering
6 how to apply machine learning algorithms to scale.

7 After a few years in industry, I decided to kind
8 of like steer my career towards civil service, which
9 is how I ended up at the SEC. And at the SEC, my
10 assessment really is that I would say that we're kind
11 of at the tail end of the kind of proof of concept and
12 kind of prototyping phase for some of our AI programs.

13 It's really exciting because some of the more
14 successful programs we're seeing a move to systemize
15 and rapidly deploy these systems at the enterprise
16 scale. Furthermore, our infrastructure both in terms
17 of compute and also data storage, it's rapidly
18 maturing.

19 At the SEC, we use artificial intelligence
20 technologies for a lot of different applications, and
21 hence, it was really important for us to kind of
22 carefully think about these applications and
23 categorize them.

24 So we've come up with two course buckets, really
25 -- and it ties to really the algorithm's potential to

1 harm the targets or sort the consequence levels, if
2 you will. So I'm going to discuss those two levels,
3 so going to talk about the high consequence bucket and
4 the low consequence bucket.

5 With regards to high consequence type models,
6 think of surveillance and risk assessment workflows,
7 so how we find insider traders or us applying risk
8 assessment algorithms to our registrants.

9 Obviously if we were careless in how we deployed
10 AI models, so if the AI had some implicit bias, we
11 could cause a lot of reputational harm to someone.
12 And furthermore, we would cause harm for ourselves
13 because we would lose the public's trust in our
14 ability to really carefully and responsibly use these
15 technologies.

16 With high consequence level algorithms, you know,
17 and I think my colleagues will agree, it's really
18 important to have humans in the loop. I think it's
19 really irresponsible to deploy these systems end to
20 end. So you will not be seeing any time soon, right,
21 an algorithm that is the judge, jury, and executioner
22 in a lot of these cases.

23 To my knowledge, no AI system is sophisticated
24 enough to understand securities laws. So it will not
25 be supplanting our human subject matter experts any

1 time soon.

2 So for us, especially in this high consequence
3 level kind of bucket, AI is just an extra piece of
4 evidence that a human or humans might use.

5 Furthermore, it's conceivable that there are actual
6 multiple models kind of helping kind of, you know,
7 derive insights to support an outcome.

8 Okay. And finally, one thing I kind of wanted to
9 discuss is that with this particular type of task,
10 it's really important when feasible to use explainable
11 AI technologies, right? For us, it's not enough to
12 say, hey, this firm -- the model says it's high risk.
13 The lawyers at the SEC will just laugh in my face.

14 What really they need is, right, an explanation,
15 right, where should I look next to find corroborating
16 evidence. As Krista had said, however, I want to note
17 that these technologies are far from solved. In fact,
18 in the field of natural language processing,
19 interpretability is still very much an open problem.
20 Right? So we have to be careful about our use of
21 these technologies in that space. Nonetheless, we
22 should try and also evaluate to see if it's
23 appropriate.

24 Finally, I wanted to discuss, right, this low
25 consequence bucket. And so you hear about these types

1 of tasks a lot less simply because it's not really as
2 exciting. Right, we kind of are interested in stories
3 about like, you know, potential future Sky Net in the
4 making, and these tasks are not that. But I believe
5 that they're a huge opportunity area.

6 And my research staff and I are actually
7 investing a lot of time here and really -- the crux of
8 this is that rote tasks can be automated in part or
9 fully using artificial intelligence. It also is nice
10 because you don't necessarily need to worry about
11 explainable AI or even things being a black box fully
12 because these tasks, again, are fairly low level and
13 inconsequential that, you know, if you get a few
14 wrong, it's not really that big of a deal.

15 But these tasks, if solved correctly, could save
16 a lot of time. And if you roll up the time saved that
17 really amounts to millions of dollars, right, across
18 enterprise. And you know, we're not even going to
19 talk about the intangible benefits here too.

20 And so what's nice about this is that at the SEC,
21 really, we consider AI technologies as an enabler,
22 right? So we want to enable staff to do more with
23 less. And really -- so these capabilities exactly do
24 just that. And again, I wanted to push forward this
25 idea of the low consequence use cases because I

1 believe that there will be plenty in every
2 organization, and if you solve them, the resource
3 savings can be big.

4 With regards to enablers, I think having mature
5 and democratizing competition in platforms is
6 important. When I first joined the SEC, you were tied
7 -- your power was tied to your (inaudible), and that
8 is no longer the case. So right now, we have the
9 capability to self-serve fairly high-powered
10 machines, even some of them have GPUs enabled, so that
11 we could write train models.

12 Also I think just really bringing the IT division
13 to bear with collaborating with them instead of having
14 an adversarial relationship, and I could discuss more
15 on that later. I think that really transformed our
16 capabilities, and we were able to progress really
17 rapidly as a result. Thank you.

18 MR. SANFORD: Thanks very much, Marco. Really
19 insightful. I really like the distinction you drew
20 between high consequence and low consequence AI. I
21 think we all have some intuitive understanding that an
22 algorithm designed to make movie recommendations is
23 very different from one that's designed to assist with
24 sentencing in a criminal justice environment, and I
25 think that kind of paradigm is very useful to keep in

1 mind as we all think about this.

2 I wanted to come back to something Krista
3 mentioned in her opening about the level of
4 partnerships and collaboration going on and maybe
5 drill down a little bit there. Ask where, when you
6 look at the federal space and you see this community
7 coming together that's focused on trying to advance
8 machine learning and AI into mission support, where do
9 you see the greatest traction occurring for
10 partnerships or collaboration? Is it with data
11 sharing? Is it developing principles? Is it
12 actually, you know, joint application development?
13 Where is the most traction happening, and what do you
14 think are some of the prospects for future such
15 engagement?

16 MS. KINNARD: Would you like me to start?

17 MR. SANFORD: Yes, please, Krista. Yeah.

18 MS. KINNARD: That's a really great question, and
19 I think that there's a lot of collaboration that we're
20 seeing. So I also want to echo your point, Marco,
21 your calling-out of high consequence versus low
22 consequence is so important. And you're right, it's
23 not glamorous, but honestly, that low consequence AI
24 is actually where I think there is the most
25 opportunity for partnership and collaboration.

1 So you can look at this from the perspective of
2 using AI to directly meet your mission. There's a
3 whole lot of functions that happens across many
4 organizations that support mission delivery. And so
5 one of the areas that I'm seeing a lot of
6 collaboration across is those mission support areas.

7 So procurement, for example, every agency, I
8 don't care who you are, you got a procurement office.
9 Maybe they do things a little bit different, but
10 there's regulations that dictates how procurements
11 happen.

12 And so if one agency has used artificial
13 intelligence or automation to help with a part of
14 that, that's something that can be shared.

15 And then I think another big area where we're
16 seeing a lot of opportunity for partnership is in
17 policy and governance models. So it's one thing to
18 build an AI solution. But I mean, it can't exist in a
19 vacuum, right? There needs to be oversight. There
20 needs to be people looking in on it. There needs to
21 be people monitoring it saying this is continuing to
22 do what we expect it to do and meet the need of the
23 organization in a way that is providing value, right?

24 And so I know on projects that I have worked on,
25 I've heard of maybe an organization not working on

1 exactly the same project, but that project that they
2 did that was different prompted them to, I don't know,
3 make the security checklist that is really valuable
4 and continue to share the checklist, right, it's --
5 it's something that can, at a very minimum, inform
6 what we're going to do at this agency, right? Or
7 policy or governance around how are you sharing data,
8 how are you assigning ownership of different datasets,
9 of different data systems, of different models, right?

10 And so that framework for how people are setting
11 this up in their organization is absolutely something
12 that I think can -- can be shared. So I'll stop
13 there. I'm sure the rest of the panel has comments
14 and ideas around that as well.

15 MR. SANFORD: Yeah. So if I could maybe pivot to
16 Sean and ask Sean to weigh on this concept of
17 partnerships, and I think you've seen partnerships,
18 again, through two different lenses in your previous
19 role and your current. But maybe talk a little bit
20 about the nature of the relationship between
21 government entity and the private sector when it comes
22 to come of these things.

23 MR. KHOZIN: Sure. I think when it comes to data
24 science and AI, (inaudible) partnerships are a
25 critical component of really starting to understand

1 how to manage high-risk use cases but also how to
2 scale some of the existing successes.

3 So when I was at the FDA working on the
4 (inaudible) I mentioned before, that was really -- the
5 essence of that was to develop portfolio (inaudible)
6 collaboration. So we started to work with a number of
7 start-ups in this domain, and in fact, published a
8 number of papers on foundational AI research as it
9 pertains to amassing the interest of patients and
10 developing new therapies. And so partnerships are a
11 critical component of that.

12 I wish there were more partnerships, public and
13 private, around (inaudible) because I think in a lot
14 of cases, we do have expertise. We do have the
15 technical infrastructure, but the data isn't there.
16 And data sharing among federal agencies, for example,
17 NIH and the FDA, and also in the context of a
18 public/private collaboration enterprise can be highly,
19 highly beneficial.

20 FDA is engaging in several public/private
21 partnerships, and obviously, we all know -- we've all
22 heard things are happening in the context of COVID-19,
23 developing a vaccine, and (inaudible) NIH has, for
24 example, collaboration (inaudible) called active
25 consortium, and data sharing is a component of that.

1 But there's another effort to (inaudible)
2 collaborating with a nonprofit organization called
3 Project Data Sphere on developing data sharing
4 frameworks, and there's a workshop coming up, I
5 believe, in October that talks about (inaudible)
6 registries, sharing data to develop new insights in
7 rare cancers.

8 But we have to think about new ways of scaling
9 responsible sharing of data. What's quite interesting
10 is that there are creative ways that in the past
11 couple of years federal agencies are thinking about in
12 terms of enabling (inaudible) medical data in the
13 evolving (inaudible) economy.

14 One of them is the efforts that is being led by
15 the Office of the National Coordinator, basically the
16 body within NHHS that's responsible for certifying
17 electronic health records.

18 In 2016, as part of 21st Century (inaudible),
19 there was a mandate directed to (inaudible) to
20 eliminate what's called information bulking electronic
21 health records, and it was traditionally -- did not
22 share the data, even in some cases for patients. It's
23 been very hard to get their own data. And typically
24 still things are faxed around and in best-case
25 scenario, it's put on a CD-ROM.

1 Now, NC currently as part of certification will
2 be mandated in January of 2021 will mandate open APIs,
3 where anyone, any individual can gain access to their
4 own health data, and they can share it with anyone
5 they wish.

6 So putting the patients at the center and in
7 charge of having their own data sort of shared with
8 researchers, companies, the federal agencies, as they
9 choose.

10 So that is likely going to be transformative. In
11 fact, those APIs have already been incorporated into
12 the Apple health kit, which has now a new electronic
13 health record future, where you are able to extract a
14 lot of the data, your own data, in a structured
15 fashion.

16 So that -- these policy decisions in combination
17 with the organizational public/private partnerships
18 are already evolving and I believe are a critical
19 component of really extracting maximum use from
20 existing machine learning methods and to be able to
21 develop better and more nuanced and more predictive
22 and precise algorithms.

23 MR. SANFORD: Thank you, Sean. Thanks very much.
24 I wanted to turn to Marco and ask a question on the
25 partnership factor as well. Again, with how you're

1 interacting with the private sector on some of these
2 things. And also to come back to a point you made,
3 the types of internal partnerships that were required
4 within SEC, you mentioned working with your IT support
5 function on some of this. So if you could address
6 those two, I think that would help folks understand
7 some of the issues too.

8 MR. ENRIQUEZ: Sure. So at the SEC, obviously,
9 we have a lot of registrants, and a lot of them are
10 technologically advanced, and they're looking for some
11 guidance with regards to how they utilize AI. And we
12 often have to walk a really fine line. We are
13 enforcing securities laws. We don't enforce good
14 practices and best practices for AI and ML use.

15 And frankly, again, going back to what Krista
16 said, a lot of these kind of things, concepts in AI,
17 again, still an active area of research. So they're
18 not solved.

19 So it goes back to this notion of, frankly,
20 really (inaudible) for us to try to issue guidance on
21 some things that are not solved, right? But we do try
22 to engage our registrants, and we actually have
23 financial -- it's called Ben Hub (phonetic). It's our
24 financial innovation hub. So registrants can request
25 to meet with us to try to present ideas and maybe

1 solicit some feedback.

2 And so that's been actually exceptionally really
3 successful. To echo what Sean said, though, we do
4 have some barriers. So data sharing is still a huge
5 kind of undertaking. In fact, it's virtually
6 nonexistent, to my knowledge.

7 Furthermore, that's also true, you know, like
8 Krista, we tried to work with other kind of financial
9 -- or other agencies just in the financial regulatory
10 domain across the globe. And even in that space,
11 sharing data and sharing code is really difficult, and
12 so we have to kind of get the blessing of a lot of
13 lawyers in succession to even share like slides, for
14 example.

15 So I think that there's a lot of work to be done
16 and maybe common MLUs and frameworks to basically
17 facilitate just sharing of information because I think
18 it really would be to everyone's advantage -- and
19 there are a lot of common tasks that we all need to
20 solve. And that's certainly true for the financial
21 services industry.

22 With regards to our IT staff, I will say when I
23 first joined SEC, it was like guerilla data science.
24 You just had to install whatever you needed to do to
25 install it on your computer to just get the job done.

1 And because asking IT often meant waiting usually a
2 week or two, which is unacceptable, especially now for
3 those of you guys who've actually coded, you literally
4 type in a command in your computer, and it installs
5 instantaneously.

6 And so that whole two weeks of wait time is just
7 ridiculous. But you know, recently, I would say the
8 last two years or so, we started really talking to IT
9 folks and saying, look, we need to come together and
10 come up with a solution because you guys don't like
11 what's happening, which is basically people
12 circumventing your policies behind your back. We
13 don't like the long times it takes to really install,
14 quite frankly, like common components that data
15 scientists use.

16 So de facto now, we made Python, for example, the
17 primary language for scientific computing at the SEC,
18 which led to a lot of like kind of IT sanctioned
19 support. So now, I could actually type -- you know,
20 install this module on my computer and have it
21 installed instantaneously.

22 At the same time, we know it's been sanctioned by
23 IT, so they -- you know, it's safe and, you know,
24 won't violate any sort of security protocols. And I
25 think that's just a really kind of good example of us

1 just coming to the table, discussing kind of like pain
2 points, and then just trying to see, like, work in IT
3 really help alleviate those pain points.

4 And so by engaging the business more, I think we
5 have now a really great IT division that I think is
6 making a lot of big strides, again, to facilitate a
7 lot of these data science and AI and (inaudible)
8 workflows.

9 MR. SANFORD: Thanks, Marco. You know, there
10 might be some federal agencies that are actually --
11 would be envious of a two-week turnaround for
12 software. Everything's a little different depending
13 on where you are.

14 I want to come back to something Kurt mentioned
15 earlier, but first, given the time, I want to invite
16 the audience members to -- attendees, please submit
17 your questions. If you have questions, we'll try to
18 get to some of those in the last minutes of the
19 webinar.

20 And as you're doing that, as you're composing
21 your questions and sending them in, I wanted to ask
22 Kurt, just to follow up, you had talked about
23 beginning to assess the effectiveness of some of these
24 systems. Could you tell us a little bit about how
25 you're doing, what you learned from the process of

1 evaluating these types of systems?

2 MR. GLAZE: Yeah, sure. I think I can give -- to
3 start, I think it'd be helpful to have a broad
4 overview of how we've attempted to assess and ensure
5 the value and quality of what we're putting out in
6 this systems.

7 So first during the development process, we
8 engaged heavily with subject matter expert attorneys
9 in the disability program to design some of the
10 discrete decision support features that we offer
11 through the program and to validate them prior to
12 release. That's normal processes.

13 But once they're released, that then transitions
14 into the evaluation stage, as you mentioned. So some
15 of the studies -- I have not personally executed
16 these, but we have statistical staff at SSA who
17 partnered with us to study outcomes of use of our
18 decision support software in disability cases during
19 our phased rollout.

20 So initially, we were a voluntary use product, so
21 kind of a natural experiment as we trained and
22 advertised our product to staff, and they could
23 voluntarily choose to engage with Insight for a
24 period. So we had a kind of natural use and nonuse
25 group at scale, and our statisticians, for instance,

1 to measure Insight's impact on case processing
2 efficiency executed a regression analysis to evaluate
3 based on our case processing system the timeliness of
4 case closure when Insight was used -- essentially,
5 when Insight was used versus not used as part of the
6 workflow.

7 And I wish they were here because they could
8 explain these much more elegantly than I could, but
9 effectively, the result was an observed reduction in
10 case processing time both at the hearings and appeals
11 level.

12 That makes sense, to some degree, a modest
13 reduction because we do offer decision support
14 features that are designed to improve the speed with
15 which they work these cases, but we also offer quality
16 feedback, which of course necessarily entails taking
17 another look, looking back at your work.

18 But net, we saw reductions. So that regression -
19 - and it takes a lot of data, a lot of voluntary use
20 data, a lot of nonuse data, and a lot of time to
21 partner with folks who aren't necessarily familiar
22 with your product and how it works and how it fits
23 into the business process. So those are large
24 studies.

25 In terms of the quality of decisions, some of the

1 studies we engaged in were evaluating the frequency of
2 Insight-covered quality areas, the frequency of
3 quality issues in those Insight-covered domains in
4 issued decisions, again, where Insight was used or not
5 used kind of as a natural experiment by our staff.
6 And again, we saw consistent reductions in Insight-
7 covered quality issues in final work products coming
8 out of our disability adjudicators at both levels.

9 We also conducted some in-depth manual case
10 studies of specific quality flags that Insight rolls
11 out where we had a team of attorneys study actual
12 cases where we raised a given quality flag and
13 actually have eyeballs on all work products, you know,
14 including manual kind of case analyses that were
15 written by our attorneys, basically on an across the
16 board look at everything they possibly touched to see
17 how they reacted. Did they catch the issue more or
18 less with Insight's help? How did it affect their
19 ultimate work products in ways that are very difficult
20 to measure through existing structured data sources
21 about our case work?

22 And again, in both of those, we did see
23 improvements in the recognition of quality issues with
24 Insight as an intervention in cases where it raised
25 that -- those flags.

1 So those are some of the studies, and this is,
2 you know, in addition to, you know, many engagements
3 with our users that also go to, as Marco said, kind of
4 intangible value of these products, which are things
5 like the dignitary interests of our claimants, the
6 workflow experience of our staff working these cases.
7 Like, what is their day to day experience with an
8 assistive tool like Insight versus not?

9 And so we do conduct outreach. We've previously
10 conducted surveys, all sorts of engagement mechanisms
11 to try and get a sense for the reception of our staff
12 and their perception of its value. And so far, we've
13 been happy that the feedback has been positive,
14 generally. So those are some of the ways.

15 MR. SANFORD: Thanks, Kurt. One quick follow-up
16 from the audience to you, Kurt, and then I've got
17 another audience question I think we have time for to
18 the whole group. The question to you, Kurt, was some
19 curiosity about what does the Insight tool actually
20 return back to the user. Is it a reading list of
21 things to look at. Is it actually marking up the
22 decision? Is it making alternative text
23 recommendations? What sort of actual product is it
24 delivering back as its recommendations?

25 MR. GLAZE: That's a very good questions. I wish

1 I had visuals because this might be a little easier.
2 But it's a web application that essentially when they
3 kick off or access Insight, it pops up essentially a
4 web browser page that displays quality feedback as
5 line items for the user to review.

6 And I should underscore again that Insight, as a
7 decision support product, is never the final arbiter
8 of any element of a disability claim's adjudication.
9 It is always an advisory service, much like if you had
10 a personal assistant helping you with a case who
11 looked over your draft before you moved it forward,
12 and they offered some observations about what they
13 saw, that this exactly what Insight is doing except at
14 scale.

15 And so it pops up a web application. We provide
16 any quality feedback we have to offer, and there may
17 be none. In something like 45 to 50 percent of cases
18 that are sent, we really don't have any substantive
19 quality feedback to provide to the user. Their work
20 product seems fine as far as Insight can tell.

21 We conduct about in the 30s about 30 discrete,
22 specific analyses of elements of quality. So we are
23 by no means comprehensive in our analysis.

24 But anyway, our feedback is provided to
25 adjudicators, and it's really a jumping-off point for

1 further analysis by that adjudicator. They can engage
2 with it. They can think -- you know, they can agree,
3 disagree, think that they've covered the issue and
4 rationale or that it's covered by the facts of the
5 case. They can change the decisional language to
6 adapt to the quality issues cited to hopefully
7 remediate it before issuance.

8 But that's where we fit in to the workflow.

9 MR. SANFORD: That's great. The last question to
10 the group and actually I'm wondering, Krista, maybe
11 you can just answer this quickly, if you can, we got a
12 question from the audience how are agencies addressing
13 or planning to address ethics in AI? Is there -- are
14 there efforts specific on the ethics question?

15 MS. KINNARD: Yes, but that's a great question,
16 and the answer is yes. So again, ethics is something
17 that's not solved, right? There is not any one single
18 person who says I know all the things about AI ethics
19 and responsible AI implementation.

20 There's many different groups thinking about
21 this, and in fact, I think every agency who is
22 thinking about AI is starting to think about this. So
23 we actually have a working group as part of our
24 community practice specifically focused on this.

25 We are not policy makers, right, we do not create

1 frameworks, we do not create policy. But we are
2 creating a space where federal government can come
3 together and share the resources they have, have those
4 tough conversations, and learn from folks who may be a
5 little bit further down the line, for example, the DoD
6 released their ethics principles. You've got other
7 organizations and groups of organizations starting to
8 release their ethics principles. So you've got a lot
9 of folks thinking about this.

10 Is there a go-to framework for how to solve AI
11 ethics and responsible AI for your organization? No.
12 But there are a lot of smart people thinking about it.

13 MR. SANFORD: Great. Thank you, Krista, and with
14 that, I see we're at time. I want to thank all of the
15 attendees at today's webinar, and in general, I want
16 to thank ACUS and the Institute for Technology, Law,
17 and Policy at Georgetown for this symposium series. I
18 think we had some fantastic insights from our
19 panelists today. I appreciate their time and wish
20 everyone well in their machine learning and AI
21 journey. Thank you very much.

22 (End of audio recording.)

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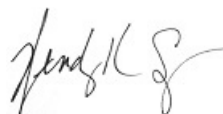
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