## Data and AI - What's Happening

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## Artificial Intelligence, Machine Learning \& Deep Learning



Training Computers instead of Programming Them

## 2019 Turing Award

## 'Godfathers of AI' honored with Turing Award, the Nobel Prize of computing

Yoshua Bengio, Geoffrey Hinton, and Yann LeCun laid the foundations for modern AI
By James Vincent \| Mar 27, 2019, 6:02am EDT
f $\underbrace{\text { share }}$


## First Tensorflow Conference



Where today's top minds bring machine learning to life
Welcome to the first TensorFlow World. Here, you'll join the pioneers and practitioners of the machine learning revolution to explore how Tensorflow is powering everything from data centers to edge devices, and environmental conservation to advanced healthcare. TensorFlow is evolving quickly. Join us, and grow with it.

The call for speakers is now open
Are you a data scientist, engineer, designer, developer, machine learning practitioner, or product manager leveraging TensorFlow to help transform your company? Are you an executive, CTO, or innovator trying to navinate how machine learnino can imnact vour business? We want to hear about vour aroundbreakino research.

$$
2018 \text { - Community Days }
$$



Great R \& @TensorFlow hacking session at \#TensorFlowDay Hacking room @oscon rladies.org followed by superb talk from @mmmpork \& @gdequeiroz - they suggested going to IBM booth to participate in @CallForCode - \& joining RLadies meetup group rladies.org
O 15 7:43 PM - Jul 17, 2018

## Center for Open Source Data and AI Technologies

Code - Build and improve practical frameworks to enable more developers to realize immediate value (e.g. FfDL, MAX, Tensorflow, Jupyter, Spark)

Content - Showcase solutions to complex and real world AI problems

Community - Bring developers and data scientists to engage with IBM (e.g. MAX)

## CODE

## CODAIT

codait (French) = coder/coded

Improving Enterprise AI lifecycle in Open Source


## The AI Ladder

A prescriptive approach to accelerating the journey to AI


AI

## INFUSE - Operationalize AI with trust and transparency

ANALYZE - Scale insights with Al everywhere

ORGANIZE - Create a trusted analytics foundation

COLLECT - Make data simple and accessible

Data of every type
regardless of where it lives


## Eras of Computing

AI Systems learn and interact naturally with people to amplify what either humans or machines could do on their own. They help us solve problems by penetrating the complexity of Big Data.


AI is "the" approach to exploit that resource

## AI Applications



## The Emergence of LeaderBoards in AI

Leaderboards - such as those in Kaggle - the home of data science contests that utilize open tech \& datasets for predictive modeling - resulting in:

- Ranking of data scientists world-wide
- Fresh datasets, data science models, methods, and education - all in open source
- Coursera class https://www.coursera.org/learn/competitive-datascience/home/welcome
- Companies (that sponsor datasets and contests) who benefit through:
- Recruitment of great employees ; Eminence of own employees; Excellent publicity ; Better understanding of what can be done with their data, Being part of a global AI conversation around open technologies


## Sample contest:



## Women in Data Science Datathon Feb 2018

Xi Lui and Ye Wang, Worcester Polytechnic Institute's data science graduate program, beat 230 teams composed of students, faculty, and professional data scientists from 26 countries.

IBM had 12 wonderful teams in the contest (more than any other institution) - the highest ranked was at 7
The contest goal was to predict if a person is male or female by examining the responses the people gave to some questions.

Kaggle: "a way to organize the brainpower of the world's most talented data scientists and make it accessible to organizations of every size" - Hal Varian, Google

| 47 | 88 | (6] | Computer says no | joined 5 years ago | 3 | 09 | - 5 | 53,959 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 48 | $g_{8}^{9}$ | - | gmobaz | joined 6 years ago | (3) | O 4 | 09 | 53,849 |
| 49 | 第 | \% | fakeplastictrees | joined 4 years ago | -6 | O 8 | C1 | 53,395 |
| 50 | \% ${ }^{\text {g }}$ | [3] | Alexey Noskov | joined 3 years ago | -6 | O 11 | 01 | 53,333 |
| 51 | \% | 2 | CPMP | joined 5 years ago | -3 | 07 | (2) | 53,022 |
| 52 | 888 | $\cdots$ | Heng CherKeng | joined 5 years ago | -2 | O 3 | 0 | 52,461 |
| 53 | \% | सि | Qingchen | joined 6 years ago | 09 | O9 | 06 | 52,024 |
| 54 | go\% | 2 | David | isined a vear aco | (2) | al | Q 3 | 51607 |

## Community Data License Agreement

There are two CDLA license agreements:

- "Sharing" based on a form of copyleft designed to encourage recipients to participate in reciprocal sharing of data
- "Permissive" an approach similar to permissive open source licenses (e.g. Apache, BSD or MIT) where recipients are not required to share any changes

http://opendefinition.org/licenses/
Candidate users of CDLA:

Communities training AI and ML systems
Public-private infrastructure (e.g. data on traffic) Researchers
Companies with mutual interests in sharing data

## More on AI



Watson Visual Recognition
Natural Language Understanding Watson Speech to Text Watson Text to Speech

Watson Visual Recognition Natural Language Classifier Watson Speech to Text

Watson Studio (Mostly open source)
Watson Machine Learning
Deep Learning as a Service
(FfDL is open source project)

## Three approaches for building AI Models



## Building AI is still hard



## Challenges of Building and deploying AI Models Today

## Al Models



## Model Asset Exchange <br> https://developer.ibm.com/exchanges/models/all/

Deployable | Facial Recognition

## Facial Emotion

Classifier

Detect faces in an image and predict the emotional state of each person

View model »

Deployable
mage-to-Image Translation 0
Transformation
Image Completer

Recognize and extract faces in an image and complete the corrupted portions.

View model»

## Deployable

| Object Detection In Images
Human Pose Estimator

Detect humans in an image and estimate the pose for each person.

View model»

Deptoyable
I Named Entity Recognition
Named Entity Tagger

Locate and tag named entities in text.

View model »

## Storing, Sharing, Reusing, Composing and Deploying Machine Learning Models



## Al Engineering: An emerging discipline

| Data handling tools |  |  |  |
| :--- | :--- | :--- | :---: |
| Image/ ()) Audio Text <br> Video    | Language |  |  |

DLaaS Cloud Platform \&
Access to Frameworks
theano Cencer

> Visualization \& HumanComputer interaction


## Al Benchmarks

DeepBench : This benchmark targets low-level operations that are fundamental to deep learning, such as matrix-multiplication, convolutions, and communications, and aims to identify the most appropriate hardware but the benchmark does not consider time-to-accuracy.

TensorFlow : The TensorFlow performance benchmarks are similar to DeepBench, in that they identify the most appropriate hardware, but not time-toaccuracy currently. They are also tied to the TensorFlow Framework.

DAWNBench : DAWNBench allows different deep learning methods to be compared by running a number of competitions. It was the first major benchmark suite to examine end-to-end deep learning training and inference. It does not address data preparation and hyper-parameter optimization work.

MLPerf : MLPerf defines the primary metric as the wall clock time to train a model to a target quality, often hours or days. The target quality is based on the current state of the art publication results, less a small delta to allow for run-to-run variance. MLPerf does not address hyper-parameter optimization nor data preparation.

The MLPerf Closed Model Division specifies the model to be used and restricts the values of the hyper parameters (batch size, learning rate, etc.) which can be tuned in an attempt to create a fair and balance comparison of the hardware and software systems.

The MLPerf Open Model Division, only requires that same task must be achieved using the same data, but provides fewer restrictions

## AI is now used in many high-stakes decision making applications



Credit
Employment
Admission
Sentencing

## What does it take to trust a decision made by a machine?

(Other than that it is 99\% accurate)


Is it fair?


Is it easy to understand?


Did anyone tamper with it?


Is it accountable?

## IBM Vision for Trusted AI

## Pittars of trust, woven into the lifecycle of an AI application



FAIRNESS


EXPLAINABILITY


ROBUSTNESS


ASSURANCE

## AI learns whatever the data teaches it

## Image Search



Language Translation


Google Translate: English / Turkish

Chatbot Interactions


The Twitter profile picture of Tay
Microsoft Tay chatted racist and xenophobic epithets learned from interacting with people

## Unwanted bias and algorithmic fairness

Machine learning, by its very nature, is always a form of statistical discrimination


Discrimination becomes objectionable when it places certain privileged groups at systematic advantage and certain unprivileged groups at systematic disadvantage.

Illegal in certain contexts:
e.g. Equal Credit Opportunity, The Equal Pay Act, The Americans With Disabilities Act,
... but not well understood in others.

Unwanted bias in training data yields models that scale the bias out

- Prejudice in labels,
- undersampling or oversampling,
- ... but bias can creep in due to incorrect model build, selection or deployment.


## AI Fairness 360

An extensible toolkit for detecting, understanding, and mitigating unwanted algorithmic bias

| IBM Research Trusted | Home | Demo | Resources | mmunity |
| :---: | :---: | :---: | :---: | :---: |

## AI Fairness 360 Open Source Toolkit

This extensible open source toolkit can help you examine, report, and mitigate discrimination and bias in machine learning models throughout the AI application lifecycle. Containing over 30 fairness metrics and 9 state-of-the-art bias mitigation algorithms developed by the research community, it is designed to translate algorithmic research from the lab into the actual practice of domains as wideranging as finance, human capital management, healthcare, and education. We invite you to use it and improve it.

```
API Docs / Get Code 
```

Not sure what to do first? Start here!

| Read More | Try a Web Demo | Watch a Video | Read a paper | Use Tutorials | Ask a Question | View Notebooks | Contribute |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Learn more about fairness and bias mitigation concepts, terminology, and tools before you begin. | Step through the process of checking and remediating bias in an interactive web demo that shows a sample of capabilities available in this toolkit. | Watch a video to learn more about AI Fairness 360. | Read a paper describing how we designed AI Fairness 360. | Step through a set of indepth examples that introduces developers to code that checks and mitigates bias in different industry and application domains. | Join our AIF360 Slack Channel to ask questions, make comments and tell stories about how you use the toolkit. | Open a directory of Jupyter Notebooks in GitHub that provide working examples of bias detection and mitigation in sample datasets. Then share your own notebooks! | You can add new metrics and algorithms in GitHub. Share Jupyter notebooks showcasing how you have examined and mitigated bias in your machine learning application. |
| $\rightarrow$ | $\rightarrow$ | $\rightarrow$ | $\rightarrow$ | $\rightarrow$ | $\rightarrow$ | $\rightarrow$ | $\rightarrow$ |

Learn how to put this toolkit to work for your application or industry problem. Try these tutorials.

| Credit Scoring | Medical | Gender Bias in |
| :--- | :--- | :--- |
| See how to detect and <br> mitigate age bias in | Expenditure | Face Images |
| predictions of credit- |  |  |
| worthiness using the German |  |  |
| Credit dataset. |  |  |$\quad$| See how to detect and |
| :--- |
| mitigate racial bias in a care |
| management scenario using |
| Medical Expenditure Panel |
| Survey data. |$\quad$| See how to detect and |
| :--- |
| mitigate bias in automatic |
| gender classification of face |
| images. |

Web experience: http://aif360.mybluemix.net/ Code: https://github.com/IBM/AIF360
Paper: https://arxiv.org/abs/1810.01943

## AI Fairness 360:

## - 30+ fairness

metrics/checkers

- 10 bias "mitigators"
- industry tutorials

Extensible, (e.g. scikit-learn's fit/predict paradigm)

Designed to translate new research from the lab to industry practitioners

## Differentiation

Comprehensive bias mitigation toolbox (including unique algorithms from IBM Research)

Several metrics and algorithms that have no available implementations elsewhere

AI Fairness 360 Open Source Toolkit
This extensible open source toolkit can help you examine, report, and mitigate discrimination and bias in machine learning models throughout the AI application lifecycle. Containing over 70 fairness metrics and 10 state-of-the-art bias mitigation algorithms developed by the research community, it is the-art bias mitigation algorithms developed by the research community,
designed to translate algorithmic research from the lab into the actual designed to translate algorithmic research from the lab into the actual
practice of domains as wide-ranging as finance, human capital management, practice of domains as wide-ranging as finance, human capital mana
healthcare, and education. We invite you to use it and improve it.

## API Docs

## Get Code

Not sure what to do first? Start here!

| Read More | Try a Web Demo | Watch a Video |
| :--- | :--- | :--- |
| Learn more about fairness <br> and bias mitigation concepts, <br> terminology, and tools before <br> you begin. | Step through the process of <br> checking and remediating <br> bias in an interactive web <br> demo that shows a sample of <br> capabilities available in this <br> toolkit. | Watch a video to learn more <br> about AI Fairness 360. |
| $\rightarrow$ | $\rightarrow$ | $\rightarrow$ |
| Read a paper | Use Tutorials |  |
| Read a paper describing how <br> we designed AI Fairness <br> 360. | Step through a set of in- <br> depth examples that <br> introduces developers to <br> code that checks and <br> mitigates bias in different <br> industry and application <br> domains. | Ask a Question |
| Join our AIF360 slack <br> Channel to ask questions, <br> make comments and tell <br> stories about how you use <br> the toolkit. |  |  |
| $\rightarrow$ | $\rightarrow$ | $\rightarrow$ |

aif360.mybluemix.net/

## AI Models are vulnerable

Adversarial<br>Robustness Toolkit

The most
comprehensive toolkit for "attacking" and defending AI


## IBM created ART, an open-source adversarial robustness toolkit

## Adversarial Robustness

- Metrics
- Adversarial Sample Detection
- Input Preprocessing

Model Robustness for Al DevOps

- Model Hardening

Model Theft

- Prevention of theft via APIs
- Detection of model theft attacks
- Deterring theft through model watermarking

Model and Data Privacy

- Provable privacy guarantees for training data (local differential privacy)
- Secure federated learning

Poisoning Attacks

- Detect poisoned training and models
- Poison can degrade performance or insert backdoors
- Develop ART as platform agnostic library
- Modular framework to evaluate robustness, generate adversarial samples, and harden models
- Integration into IBM offerings to build secure model building pipelines

IBM ART
Adversarial Robustness Toolbox
a.k.a. Nemesis

## Adversarial Robustness Toolbox (ART)

External: https://github.com/IBM/adversarial-robustness-toolbox

- Python library, 7 K lines of code
- State-of-the-art attacks, defences and robustness metrics

| Load ART modules | from keras.datasets import mnist from keras.models import load_model |
| :---: | :---: |
|  | from art.attacks import CarliniL2Attack from art.classifier import KerasClassifier from art.metrics import loss_sensitivity |
|  | $\begin{aligned} & \text { \# Load data } \\ & \text { (_, _), (x_test, y_test) = mnist. load_data() } \end{aligned}$ |
| Load classifier model (Keras, TF, PyTorch etc) | \# Load model and build classifier model = load_model('my_favorite_keras_model.h5') classifier $=-$ KerasClassifier( $(0,1)$, model) |
| Perform attack | \# Perform attack <br> attack $=$ CarliniL2Attack(classifier) <br> adv_x_test $=$ attack.generate(x_test) |
| Evaluate robustness | \# Compute metrics on model robustness print(loss_sensitivity(classifier, x_test)) |

Open-source release @ RSA 2018:

~3.5K+ views of IBM blogs
~ 100+ news outlets covering release
~ $1.3 \mathrm{M}+$ Social Media potential impressions
~5K+ views of GitHub repo
^silicon^NGLE
Attackers can fool Al programs. Here's how developers can fight back

r zames kobielus
PDATED 00.53 EST . 22 APRIL 2018
IBM launches open-source library for securing Al systems
The framework-agnostic software library contains attacks, defenses, and benchmarks for securing artificial

IBM ENTWICKELT WERKZEUGE GEGEN HACKERANGRIFFE DURCH "BÖSE" KI

Выпущена Adversarial Robustness Toolbox, открытая библиотека от IBM для защиты ии

Adversarial Robustness Toolbox: IBM propose une
boite à outils open source pour sécuriser l'intelligence
artificielle
Par: fredericmazue |jeu, 19/04/2018-12.29
23-04-2018 | door: Witold Kepinski
N intelligence artificielle, attaque contradictoire

IBM Adversarial Robustness Toolbox beschermt tegen kwaadaardige AI

## Trust in AI Systems

## Factsheets for Al Services

Concerns about safety, transparency, and bias in AI are widespread, and it is easy to see how they erode trust in these systems Part of the problem is a lack of standard practices to document how an Al service was created, tested, trained, deployed, and evaluated; how it should operate; and how it should (and should not) be used. To address this need, my colleagues and I recently proposed the concept of factsheets for AI services. In our paper [1], we argue that a Supplier's Declaration of Conformity (SDoC, or factsheet, for short) be completed and voluntarily released by Al service developers and providers to increase the transparency of their services and engender trust in them. Like nutrition labels for foods or information sheets for appliances, factsheets for A services would provide information about the product's important characteristics. Standardizing and publicizing this information is key to building trust in Al services across the industry.


## Adversarial Attack Example - Pandas \& Capuchins



- Perturb model inputs with crafted noise
- Model fails to recognize input correctly
- Attack undetectable by humans
- Random noise does not work.


## AI Fairness 360

The AI Fairness 360 toolkit (AIF360) is an open source software toolkit that can help detect and remove bias in machine learning models.
x

Get the code

The AI Fairness 360 toolkit (AIF360) is an open source software toolkit that can help detect and remove bias in machine learning models. It enables developers to use state-of-the-art algorithms to regularly check for unwanted biases from entering their machine learning pipeline and to mitigate any biases that are discovered.

## AI Ethics https://www.ibm.com/blogs/policy/francesca-rossi-

 ai/
## https://www.ibm.com/watson/assets/duo/pdf/everydayethics.pdf

 world's most enduring problems, from discovering insights in data to treat disease and predict global weather events to managing the global economy and bringing populations out of poverty.At IBM, we fully subscribe to an ethical approach to AI and have stated our commitments on different ethicsrelated issues in our Principles for Trust and Transparency. They include: developing AI to augment human intelligence rather than replacing it; providing transparency and explainability in AI systems; and detecting
 and mitigating AI bias both in data and models. Our Principles also ensure that clients retain ownership and control of their data in the AI systems we deploy.

The goal of the European Commission's AI Expert Group is ambitious in vision, yet pragmatic and hopefully impactful in achieving results. The first task will be to build AI ethics guidelines, and there will be a broad spectrum of subjects to

## Watson Studio

## IBM Watson Studio

71 Reviews-G2 Crowd
Build and train AI \& machine learning models, prepare and analyze data

- all in a flexible, hybrid cloud environment

Start on cloud for free

## Explore the product tour

$\rightarrow$ Learn about Watson Studio Desktop

## Code Patterns

## https://developer.ibm.com/patterns/

## Code Patterns

Technologies - Industries . Deployment Models ~ Sort by Newest First •

Analytics x


## IBM AI Learning \& Certification:

 AI Literacy for ALLSharing our deep AI knowledge \& experience from working with hundreds of enterprise clients.

Now Live:
Online learning with our AI Learning Catalog
Custom, in-lab AI Learning Experiences Coming Soon:
IBM AI Certification for the end-to-end enterprise AI workflow.

Invest in your future. Start your AI Learning now.
http://community.ibm.com/aiskills

## Get involved

## Call for Code 2019



## How can you participate?

Developers register for the challenge, get started building applications that will save lives.
www.developer.ibm.com/callforcode
Support Call for Code:

- Host a day for your organization
- Provide promotional support for the initiative
- Donate in-kind: charitable donations, offer a VC pitch to the winning team or donate your technology
https://callforcode.org/become-a-supporter/
Sponsor, show your full support with a sponsorship.
https://callforcode.org/become-a-sponsor/

Visit www.developer.ibm.com/callforcode

Backup

## Data \& AI : What's Happening

Talk Summary : This session will provide a brief overview of what's happening with AI with a particular emphasis on data - and then provide a summary of the IBM offerings and products that support AI and Data Science

Bio : Susan Malaika is Senior Technical Staff in the Cognitive Applications group in IBM focusing on open source for Data \& AI, Susan also leads a tech community of a few hundred volunteers in the New

York area \& she loves hackathons.
For more information about Susan please see https://developer.ibm.com/opentech/category/susan-malaika/

## MAX - Model Asset Exchange

## AI is Not Magic: It's Time to Demystify and Apply

A unified, modern data fabric.

A development
environment and engine.

Human features.
AI management and exploitation.

## Participate : Registrations Opened 2019-03-25-https://callforcode.org/

## CALL FOR CODE 2019

100,000 developers from 156 nations accepted the challenge and built over 2,500 applications in 2018.
Over 60 organizations amplified the Call as supporters last year.

## IBM Corporate Service Corps

Throughout its 10 years, IBM CSC has:


Activated over 4,000 participants from 62 different countries
"The Corporate Service Corps provides an enormous growth opportunity with exposure to real-life challenges and cultural differences which can't be matched in normal work settings. It was inspiring to see the personal growth in colleagues from across the world as we strived to make genuine community impact."

CSC India Team Participant


Supported over 340 teams with more than 1,400 projects

## 

IBM Corporate Service Corps Collaborators Projects Press

$\downarrow$ Overview
$\downarrow$ Program details
$\downarrow$ Case studies

## Natural disasters are among the world's greatest challenges...

## $800,000+$

worldwide deaths attributed to earthquakes since 2010

## 25\%

coastline areas that met or surpassed record number of flood days

## 800+

confirmed tornadoes touched down in 2018

## 17 million

acres lost to wildfire in the United States in the last 2 years
22
named storms in the Eastern Pacific region this year a record


CA
volcanos considered a "very high threat" in the U.S. alone

## Agenda

- Data
- ML - Machine Learning,
- DL - Deep Learning
- AI Trust
- Join the Call for Code


## Code ｣ <br> Response

Code and Response is an IBM initiative which provides a platform to create and deploy open source technologies to tackle some of the world's biggest challenges.

Coding challenges includes Call for Code, CGIU student codeathons
Solution deployment starting with Call for Code 2018 winner Project OWL
Volunteer in disaster relief efforts with the American Red Cross \& more

Code and Response ${ }^{\text {TM }}$ is supported by NGOs, governments, global technologists, as well as the IBM Corporate Service Corps.

## CALL FOR CODE

## 100k <br> Developers <br> 156 <br> Nations <br> 2,500+ <br> Applications

Part of Code and Response ${ }^{T M}$, this annual global developer challenge is a great way to get involved. It inspires developers to create sustainable software solutions to prepare for, respond to and recover from natural disasters. www.developer.ibm.com/callforcode

The winning team receives:

- A \$200K cash prize
- Open Source Support from The Linux Foundation
- Meetings with mentors and potential investors
- Solution implementation through Code and Response ${ }^{T M}$

| Get involved | Support Call for Code, <br> and host a day |
| :--- | :--- | | Become an affiliate, donate |
| :--- |
| in-kind | and host a day

Sponsor, show your full support with a sponsorship

| Call for Code challenge opens March 25 | Project Owl Implementation April | Cause Flash <br> (UN World Health Day) April 7 | Wildfire Community Preparedness Day (+42 school event) <br> May 4 | National Hurricane Preparedness Week May 10 | Cause Flash (World Environment Day) June 5 | World Humanitarian Day <br> Aug 19 | Award Event October 13 NYC |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |

The IBM open source way
https://developer.ibm.com/open/culture/

| OpenSource BMOS |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Training <br> Open Source @ IBM Program touches | Recognition <br> We recognize our open source leaders with | Tooling <br> Our open source management tool suite is used over | Organization <br> Our Open Source Core Team includes | Consuming <br> Virtually all of our products contain open source | Contributing <br> We invest in community code \& innovation |
| $78,000$ <br> IBMers annually | $300+$ <br> cash awards annually | $30,000+$ <br> times per month | ~12 FTEs <br> sunnorting all of <br> IBM | 3000+ <br> packages reviewed every month | 1500+ <br> GitHub repos |

## Datasheets Proposal

- The machine learning community has no standardized way to document how and why a dataset was created, what information it contains, what tasks it should and should not be used for, and whether it might raise any ethical or legal concerns. To address this gap, we propose the concept of datasheets for datasets.
- In the electronics industry, it is standard to accompany every component with a datasheet providing standard operating characteristics, test results, recommended usage, and other information. Similarly, we recommend that every dataset be accompanied with a datasheet documenting its creation, composition, intended uses, maintenance, and other properties.
- Datasheets for datasets will facilitate better communication between dataset creators and users, and encourage the machine learning communityto prioritize transparency and accountability.


## Sample questions:

- Why was the dataset created? (e.g., was there a specific intended task gap that needed to be filled?)
- Who funded the creation of the dataset?
- What preprocessing/cleaning was done? (e.g., discretization or bucketing, tokenization, part-of-speechtagging, SIFT feature extraction, removal of instances)
- If it relates to people, were they told what the dataset would be used for and did they consent?If so, how? Were they provided with any mechanism to revoke their consent in the future or for certain uses?
- Will the dataset be updated? How often, by whom?


## Summary

This talk reviews the challenges and metrics for enterprise workloads, the benchmark tests that are available, and the gaps which need to be filled.

The paper, that this talk is based on, identifies the following areas as important to enterprises concerned about performance:

1. Model training performance
data labeling / preparation
time-to-accuracy
computational time / cycle

- throughput-to-accuracy


## 2. Hyper-parameter optimization performance

3. Inference runtime performance

The talk offers a summary table of the main three Al areas important to enterprises, alongside:

- Workload profile
- Important performance indicators to assess the task's efficiency

Potential technical bottlenecks to look out for that could limit the AI tasks performance delivered by a given solution.

