Data and AI – What's Happening

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Artificial Intelligence, Machine Learning & Deep Learning

Fact Sheets for Data and AI Data & Model Licenses ARTIFICIAL INTELLIGENCE A program that can sense, reason,

MACHINE LEARNING

act, and adapt

Algorithms whose performance improve as they are exposed to more data over time

TensorFlow DEEP LEARNING PyTor

Subset of machine learning in which multilayered neural networks learn from vast amounts of data

AI Trust

Models

Model Exchange

Soon, we won't program computers. We'll train them. at JASON TANK code*/

> What that means for us. BY CRUE NETZ

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





From left to right: Yann LeCun | Photo: Facebook; Geoffrey Hinton | Photo: Google; Yoshua Bengio | Photo: Botler Al

First Tensorflow Conference

2019 - Conference



Request sponsorship information

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 navigate how machine learning can impact your business? We want to hear about your groundbreaking research.

2018 – Community Days



Susan Malaika @sumalaika y

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 ♡ 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)

IBM



codait.org

codait (French) = coder/coded

https://m.interglot.com/fr/en/codait

Improving Enterprise AI lifecycle in Open Source



The AI Ladder

A prescriptive approach to accelerating the journey to AI



INFUSE – Operationalize AI with trust and transparency

ANALYZE - Scale insights with AI everywhere

ORGANIZE - Create a trusted analytics foundation

COLLECT - Make data simple and accessible





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.



Data is "the" natural resource

Big Data

AI is "the" approach to exploit that resource

AI Applications

Speech Recognition lets you go hands-free

Autonomous Vehicles detect pedestrians

> Machine Vision detects cancer early

Facial Recognition unlocks your phone

Machine Learning and Al are everywhere

> Spam Detection unclogs your Inbox

Fraud Detection protects your credit

> Recommendations help you shop faster

Chat Bots route calls quicker

The Emergence of LeaderBoards in AI

Leaderboards – such as those in <u>Kaggle</u> - 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. Kaggle: "a way to organize the brainpower of the world's most talented data scientists and make it accessible to organizations of every size" - <u>Hal Varian</u>, Google



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 <u>the responses the people gave to some</u> <u>questions</u>.

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

Current practices around sharing data vary but generally map to requirements we've dealt with in source code licensing

Open data publishers are currently using multiple approaches to open licensing data

- > Public Domain, see: <u>https://opendatacommons.org/guide</u>
 - Data.gov "Additionally, we waive copyright and related rights in the work worldwide through the CC0 1.0 Universal public domain dedication."
- > Open Source Licenses, CC BY-SA 2.0
- > Open "Data Licenses", see http://wiki.openstreetmap.org/wiki/Open_Database_License
- Canadian Government publishes data under the "Open Government Licence", see <u>http://open.canada.ca/en/open-government-licence-canada</u>
- > Some communities only ask for attribution...
 - "The CHIANTI package is freely available. If you use the package, we only ask you to appropriately acknowledge CHIANTI." (http://www.chiantidatabase.org)
- Currently difficult to understand ability to combine datasets from different licenses
- > No one has figured out a "weak copyleft" model

License	Domain	By	SA	Comments	
Creative Commons CCZero (CC0)	Content, Data	N	N	Dedicate to the Public Domain (all rights waived)	
Open Data Commons Public Domain Dedication and Licence (PDDL)	Data	N	N	Dedicate to the Public Domain (all rights waived)	
Creative Commons Attribution 4.0 (CC-BY-4.0)	Content, Data	Y	N		
Open Data Commons Attribution License (ODC-BY)	Data	Y	N	Attribution for data(bases)	
Creative Commons Attribution Share-Alike 4.0 (CC-BY-SA-4.0)	Content, Data	Y	Y		
Open Data Commons Open Database License (ODbL)	Data	Y	Y	Attribution- ShareAlike for data(bases)	

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

License Announced in November 2017 by Linux Foundation

https://www.linuxfoundation.org/press-release/2017/10/linux-foundation-debuts-community-data-license-agreement/

More on Al



Three approaches for building AI Models



Building AI is still hard

Lack of massive labeled data sets in enterprises

Lack of reliable tools for monitoring No feedback loop to improve models in situ

> No standards to inject AI No automation Concerns over security, bias, and ethics



Challenges of Building and deploying AI Models Today



Model Asset Exchange https://developer.ibm.com/exchanges/models/all/



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

dels

What is ONNX?

ONNX is a open format to represent deep learning models. With ONNX, AI developers can more easily move models between state-of-the-art tools and choose the combination that is best for them. ONNX is developed and supported by a community of partners.

	Hewlett Packard Enterprise	Tencent	(intel) Al
IBM.	MEDIATEK	arm 火	HUAWEI
📣 MathWo	orks Qual	comm @	
🚭 unity	Bai de 百度	CEVA	Oath:
Preferred Networks		skymizer	SYNOPSYS'

ONNX tutorials: import and export from frameworks

Framework / tool	Installation	Exporting to ONNX (frontend)	Importing ONNX mo (backend)
Caffe2	part of caffe2 package	Exporting	Importing
PyTorch	part of pytorch package	Exporting, Extending support	coming soon
Cognitive Toolkit (CNTK)	built-in	Exporting	Importing
Apache MXNet	part of mxnet package docs github	Exporting	Importing
Chainer	chainer/onnx-chainer	Exporting	coming soon
TensorFlow	onnx/onnx-tensorflow	Exporting	Importing [experimenta
Apple CoreML	onnx/onnx-coreml and onnx/onnxmltools	Exporting	Importing
SciKit-Learn	onnx/onnxmltools	Exporting	n/a
MLNET	built-in Convert to ONNX-ML	Exporting	n/a
Menoh	pfnet-research/menoh	n/a	Importing

Portable Format for Analytics (PFA)	
Motivation:	
What is PFA for? +	
Interactive Tutorials:	
Tutorial 1: Scoring engines	
Tutorial 2: Programming	
Tutorial 3: Data flow	
Exoplanets example	
atistical models	
terences:	
cument structure	
ro types	
ecial forms	
nction library	

What is PFA for?

Hardening a data analysis

Data analysis is not software development: a different set of best practices apply. For a large software project, one should start by designing a maintainable architecture, but for data analysis, one should start by examining the dataset in as many ways as possible. Sometimes, a simple observation in this exploratory phase dramatically changes one's analysis strategy.

The worlds of data analysis and software development clash when a poorly structured analytic procedure must be scaled up to a large production workflow. The "try anything, get feedback quickly" mindset that was an asset in the development phase leads to failures in production. As data analyses mature, they must be hardened— they must have fewer dependencies, a more maintainable structure, and they must be robust against errors.



v43 v421 v41 v461 v22 v31 v38 v21 v28 v11 Examples RFC ManagementProcess	
PMML 4.3 - General Structure	PMML4.3 Menu
THE STATE IN CONTRACT ON THE ADDRESS OF ADDR	Home
Privil uses with to represent mining indees, the solution or ine indees to eschere up an AxE schema. One or ince imming indees can be contained in a Privil, document is an AxE document with a root element or type. The general structure of a Privil document is.	Chappes
	XML Schema
Chail examinit.(F): Other Examinit.(F): Market example.(F): Analytic example.(F): Chailer example.(F): Ch	Conformance Interceptability Central Stancture Pieto Scope Interce Cata Ectorology Minrog Schema Hannes
The namespaces in the PMML Schema Riself are defined as:	Statistics
cest schemes Malian aus Mithy I.//www.edl.ceg/2001/DBCalchemes* Large climane parter Mithy I.//www.edl.ceg/2002/DBCalchemes* Large climane parter Mithy I.//www.edl.ceg/2002/DBCalchemes* Large climane parter Mithy I	Jaconolog Targets Output Functions Butt-to Functions Model Ventication Model Explanation
Note that because of the namespace declaration in its current form, PMAL cannot be mixed with content of a different namespace.	Multiple Models

Although a PMML document must be valid with respect to the PMML XSD, a document must not require a validating parser, which would load external entities. In addition to being a valid XML document, a valid PMML document must obey a number of

AI Engineering: An emerging discipline





Visualization & Human-Computer interaction









AI 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)







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Is it fair?

Is it easy to understand?

Did anyone tamper with it?

Is it accountable?

IBM Vision for Trusted AI

Pillars of trust, woven into the lifecycle of an AI application



AI learns whatever the data teaches it



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 AI

Home Demo Resources Community

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.



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 in- depth 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 tooikit.	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 show- casing how you have examined and mitigated bias in your machine learning application.
→	\rightarrow	\rightarrow	\rightarrow	\rightarrow	\rightarrow	\rightarrow	\rightarrow

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

Medical

Survey data.

Expenditure

See how to detect and

mitigate racial bias in a care

management scenario using

Medical Expenditure Panel

Cre	dit	Sco	orir	ıg
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See how to detect and mitigate age bias in predictions of creditworthiness using the German Credit dataset.

Gender Bias in Face Images

-

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

Trusted AI **AI Fairness 360:**

- 30+ fairness metrics/checkers
- ▶ 10 bias "mitigators"

industry tutorials

Differentiation

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

Several metrics and algorithms that have no available implementations elsewhere

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

Designed to translate new research from the lab to industry practitioners

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-ofthe-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 wide-ranging 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
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.
÷	→	<i>→</i>
Read a paper	Use Tutorials	Ask a Question
Read a paper describing how we designed AI Fairness 360.	Step through a set of in- depth 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.
		-

aif360.mybluemix.net/

Trusted AI: Making AI Secure

AI Models are vulnerable



Adversarial Robustness Toolkit

The most comprehensive toolkit for "attacking" and defending AI



IBM created ART, an open-source <u>a</u>dversarial <u>r</u>obustness <u>t</u>oolkit

Adversarial Robustness

- Metrics
- Adversarial Sample Detection
- Input Preprocessing
- 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

Model Robustness for AI DevOps

- 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

https://adversarial-robustness-toolbox.readthedocs.io/en/latest/



Adversarial Robustness Toolbox (ART)



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

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

from keras.datasets import mnist from keras.models import load model from art.attacks import CarliniL2Attack Load ART from art.classifier import KerasClassifier modules from art.metrics import loss sensitivity # Load data (_, _), (x test, y test) = mnist.load data() # Load model and build classifier Load classifier model = load model('my favorite keras model.h5') model (Keras, classifier = KerasClassifier((0, 1), model) TF, PyTorch etc) # Perform attack attack = CarliniL2Attack(classifier) Perform attack adv x test = attack.generate(x test) # Compute metrics on model robustness Evaluate print(loss sensitivity(classifier, x test)) robustness

Open-source release @ RSA 2018:



- ~ 3.5K+ views of IBM blogs
- ~ 100+ news outlets covering release
- ~ 1.3M+ Social Media potential impressions
- ~ 5K+ views of GitHub repo

∧ silicon∧NGLE

Attackers can fool AI programs. Here's how developers



Trust in AI Systems

Factsheets for AI 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 AI 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 AI 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 AI services would provide information about the product's important characteristics. Standardizing and publicizing this information is key to building trust in AI services across the industry.





Aleksandra Mojsilovic IBM Fellow, IBM Research

Adversarial Robustness Toolkit

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.

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.

AIF360 enables AI developers and data scientists to easily check for biases at multiple points

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

→ Learn about Watson Studio Desktop

Code Patterns

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



IBM AI Learning & Certification: AI Literacy for ALL

Sharing 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. <u>http://community.ibm.com/aiskills</u>

Get involved Call for Code 2019



How can you participate?

Developers register for the challenge, get started building applications that will save lives. <u>www.developer.ibm.com/callforcode</u>

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. <u>https://callforcode.org/become-a-sponsor/</u>

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

- MAX is a one-stop exchange for data scientists and AI developers to consume models created using their favorite machine learning engines like TensorFlow, PyTorch, and Caffe2, and provides a standardized approach to classify, annotate, and deploy these models for prediction and inferencing.
- Visit the Model Asset Exchange at: <u>https://developer.ibm.com/code/</u> <u>exchanges/models/</u>

IBM Code Model Asset Exchange

A place for developers to find and use free and open source deep learning models.

All models





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.

Will you answer the call in 2019?

Accept The 2019 Developer Challenge

Amplify The Call As A Supporter

IBM Corporate Service Corps

Throughout its 10 years, IBM CSC has:

impact."

CSC India Team Participant



Activated over 4,000 participants from 62 different countries



Deployed to 44 different countries

IBM



Supported over 340 teams with more than 1,400 projects

Q 8 E



↓ Overview

↓ 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

A Contract of the storms in the Eastern Pacific region this year – a record

18

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 r→ 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[™] is supported by NGOs, governments, global technologists, as well as the IBM Corporate Service Corps.

www.developer.ibm.com/code-and-response

CALL FOR CODE

100k1562,500+DevelopersNationsApplications

Part of Code and Response[™], 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. <u>www.developer.ibm.com/callforcode</u>

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[™]

Get involved	Support and hos	t Call for Code, t a day	Become an in-kind	come an affiliate, donateSponsor, show your fullkindsupport with a sponsors		show your full ith a sponsorship	t ship	
Call for Code challenge opens March 25	Project Owl Implementation April	Cause Flash (UN World Health Day) April 7	Wildfire Community Preparedness Day (+42 school event) May 4	National Hurricane Preparedness Week May 10	Cause Flash (World Environment Day) June 5	World Humanitarian Day Aug 19	Award Event October 13 NYC	
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The IBM open source way https://developer.ibm.com/open/culture/

OpenSource				openSource IBNOS developeribm.com/code	Image: State
Training	Recognition	Tooling	Organization	Consuming	Contributing
Open Source @ IBM Program touches	We recognize our open source leaders with	Our open source management tool suite is used over	Our Open Source Core Team includes	Virtually all of our products contain open source	We invest in community code & innovation
78,000 IBMers annually	300+ cash awards annually	30,000+ times per month	~12 FTEs supporting all of IBM	3000+ packages reviewed every month	1500+ 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 community 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 / cycles
 - throughput-to-accuracy
- 2. Hyper-parameter optimization performance
- 3. Inference runtime performance

The talk offers a summary table of the main three AI 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.