

An Introduction to NETL's Science-based AI/ML Institute

May 13, 2021



NETL's Science-based AI/ML Institute (SAMI) TENED TENED







HEADQUARTERED



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Chung Shih TECHNICAL LEAD

- SAMI was established in 2020
- Combines the strengths of NETL's energy computational scientists, data scientists, and subject matter experts with strategic partners
- SAMI is innovating solutions to today's energy challenges

Learn more about SAMI: https://edx.netl.doe.gov/sami/





Leverage science-based models, AI/ML methods, data analytics, and high-performance computing to accelerate applied technology development for clean, efficient, and affordable energy production and utilization



A few definitions...and why AI/ML Now?



WIRED



Artificial Intelligence

• "Programmed" intelligence

Machine Learning

- Supervised ML, the machine is trained, taught
- Unsupervised ML, the machine learns on it's own (Google's cat video experiment)

Big Data

 Large volumes, variety, variability, velocity of data

Big Data Computing

 Computing engineering & systems to handle big data







storage (Information)

Statistics Big Data (Information)

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Statistics (Information)

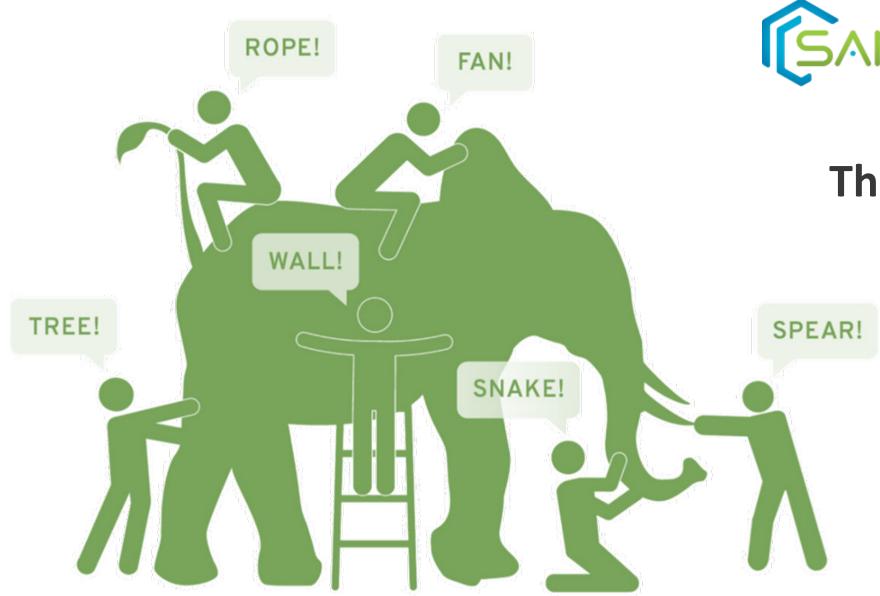
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The Opportunity

AI/ML offers new ways to explore old problems



AI/ML use in daily life has sky rocketed







Auto-complete



Virtual Assistant



Online Shopping







Suggester Models for TV Shows



Driving Route Guidance



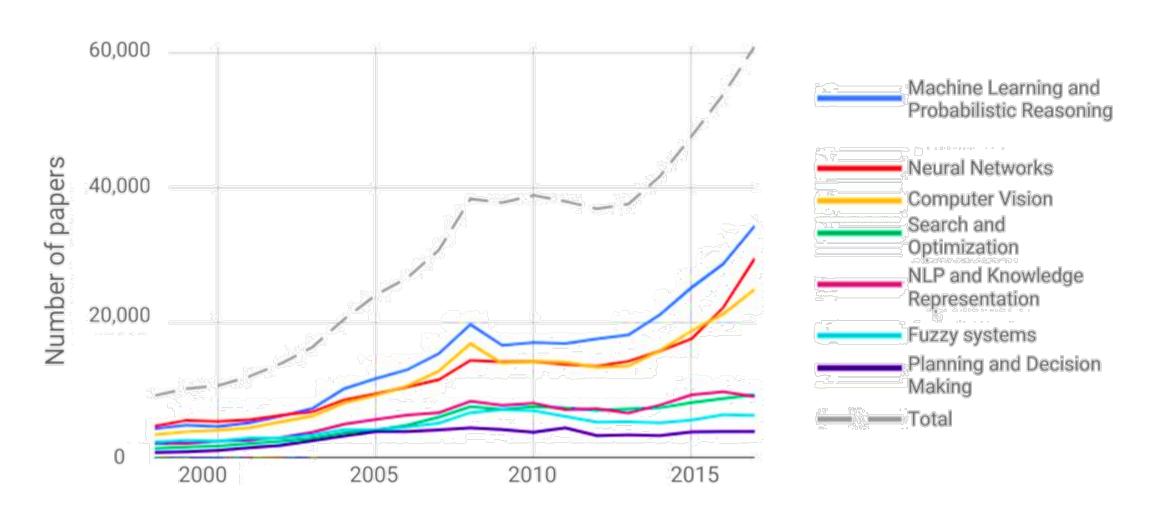
Self-driving Cars



AI/ML R&D Breakthroughs are Accelerating Too SAMI | NE



Number of Al Papers on Scopus (1998-2017), Source: Elsevier





3 MAJOR CHALLENGES in Applying AI/ML



SCIENTIFIC DISCOVERY IS DEPENDENT ON DATA

Data offers significant opportunities to drive advanced AI/ML breakthroughs, but research is often impeded by data access, quality, formats, gaps, multi-source, multi-scale, and integration challenges.



COMPLEX AI/ML MODELS CAN BE DIFFICULT TO UNDERSTAND

The logic of complex AI/ML, such as deep learning models, can be difficult to understand. It is critical to have explainable and trustworthy AI.

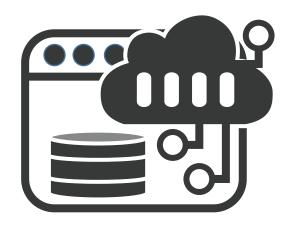


INCONSISTENT PRACTICES MAKE EVALUATION DIFFICULT

Even with training, validation, and testing processes, unnoticed incorrect workflow details can still significantly impact the results.



Injecting science and physics to guide AI/ML



SMART DATA PLATFORM (EDX++)

R&D data & knowledge refinery with connectivity to computing resources



SCIENCE-BASED AI/ML MODELING

Address previously unanswerable problems



FOUNDATIONAL STANDARDS & PLATFORMS

Establish responsible use of AI/ML



Connecting data with scientific computing Driving next-gen R&D...



Inform

Analyze & Optimize

Integrate & Label

Explore & Transform

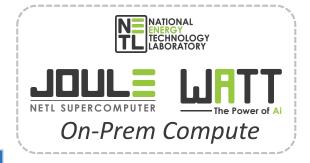
Move & Store

Discover & Collect

EDX++ FRAMEWORK







...ensuring compliance with Federal/DOE regulations

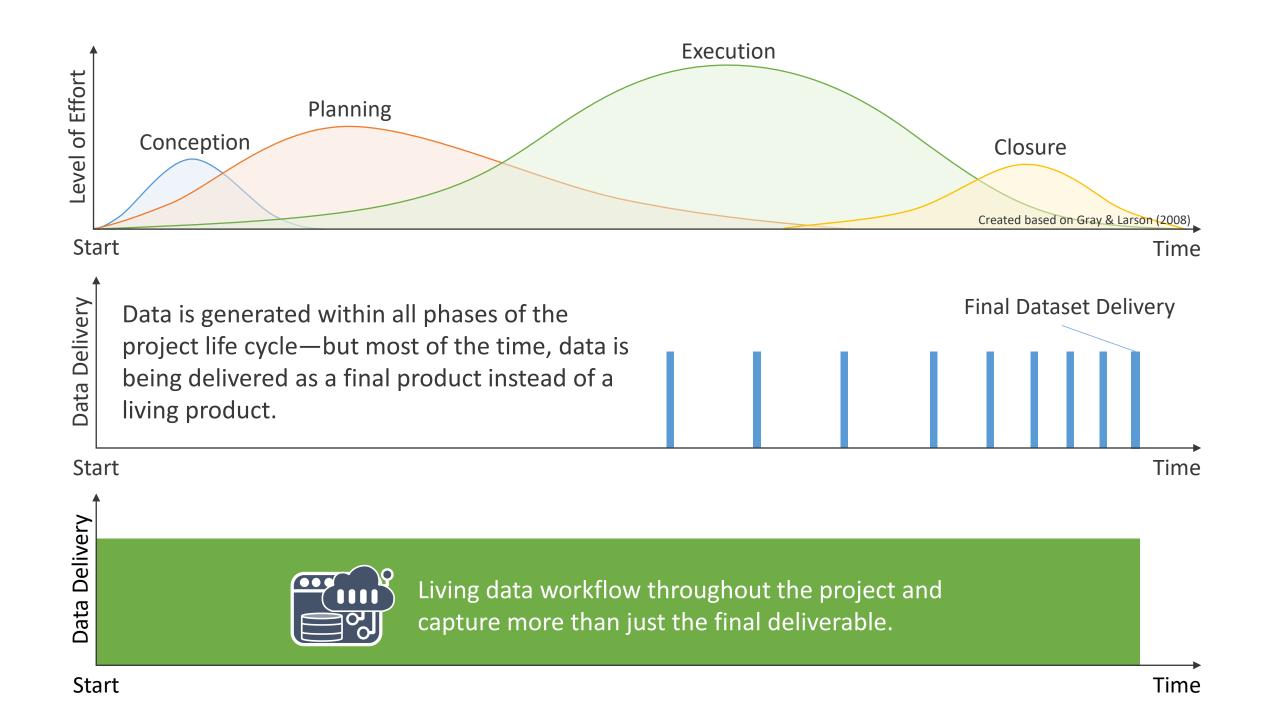


...ensuring preservation and access to DOE FE knowledge and data resources



Big Data





EDX is more than just a data repository







Search

Submissions within the public search on EDX provide access to many forms of information including but not limited to presentations, publications, tools, and data.



Submissions within the public search on EDX can be sorted spatially, by keyword, and file **format** connecting users to the appropriate data and information quickly and efficiently.



Submissions within the public search on EDX can be clustered into Groups of related data. Some popular EDX Groups include the Kimberlina Data Group, Appalachian Basin Data Group, and various RCSPs.







EDX Tools provide access to, management of, and interaction with data through a collection of tools including SmartSearch, Natcarb Viewer 2.0, CSIL and NRAP Tools.



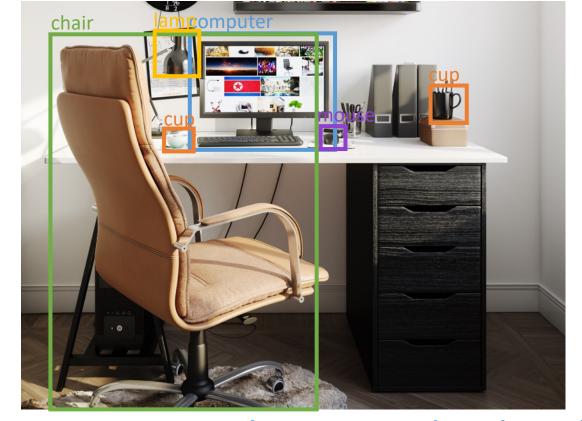
EDX Tools like Geocube, Natcarb Viewer, and Blosom allow users to find, sort, visualize, and interact with geospatial data.

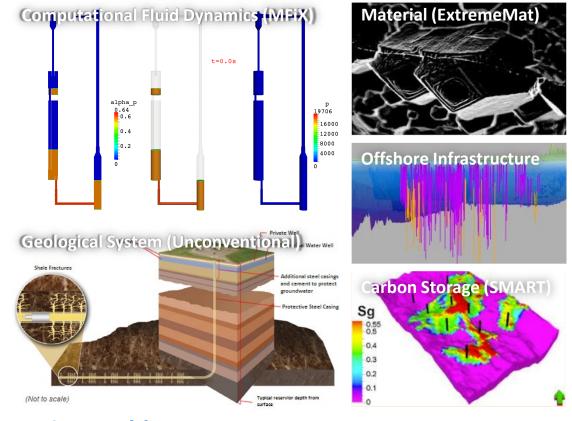
Visualize



EDX Tools provide visualization of data through various tools including ParaView, Papaya, and RokData (coming soon).







There are two ends on the machine learning problem spectrum

Data Driven

- Object Detection and Classification
- Spam Identification
- Recommendation Systems
- Autonomous Driving
- Natural Language Processing
- More...

With Scientific and Physics Prior Knowledge

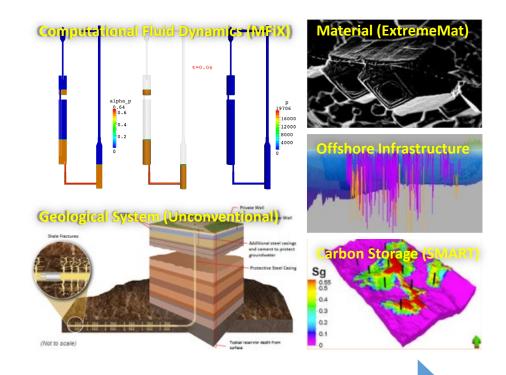
- Computational Science and Engineering
- Computational Fluid Dynamics
- Geological and Environmental Systems
- Material Engineering and Manufacturing
- Energy Conversion Engineering
- More...



Challenges with this end of the spectrum







With Scientific and Physics Prior Knowledge

- Data acquisition can be costly and timeconsuming, resulting in limited available data for AI/ML training
- Complex and high-dimensional space (with many variables)
- Multi-scale tasks
 (e.g., time, geospatial, atom-to-process)
- Complex AI/ML models can be difficult to understand and thus trust





Science-based AI/ML Modeling

Address previously unanswerable problems

Incorporate Science/Physics into AI/ML



Generate synthetic data to overcome the challenge of limited observation data based on scientific simulations



Enhance AI/ML loss functions with scientific knowledge

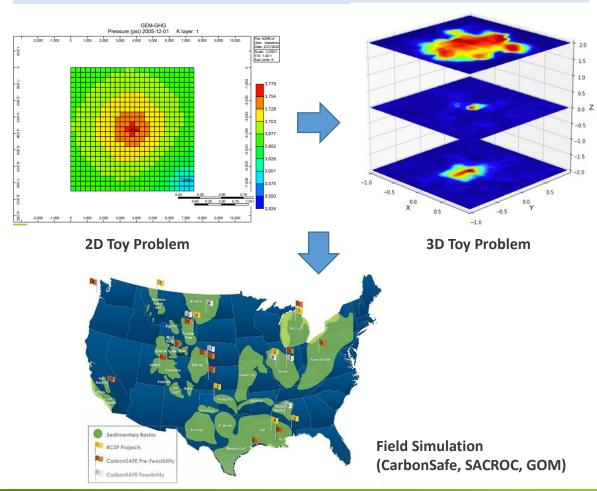
Loss = Prediction Data Fitness + Physics Fitness

Benefits

- Use less data to guide model training
- With the same amount of data, improve the overall accuracy
- Increase the confidence with the results

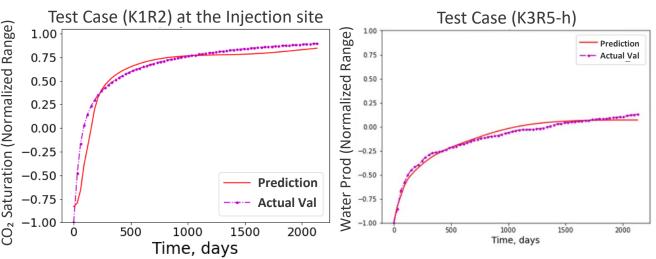
Part of the SMART Initiative to build a learning system to predict CO₂ storage behaviors

Dataset Dev w/ Scientific Simulations



Approach & Preliminary Results

- Designed deep learning models to capture the spatiotemporal changes of CO₂ saturation, pressure, and water production
- Leveraged NETL's High Performance Computers (Joule and Watt)
- Constraints and regularizations are essential
- Started to incorporate physics into models





AI/ML Combined with Science-based Models is **Enabling Breakthroughs in Applied Energy R&D**



Power Plant Resiliency and Reliability



Advanced Materials Discovery



Spatial and Subsurface Analysis, Prediction, and Visualization



































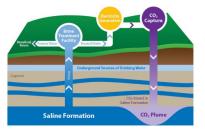


NETL supported with AI/ML capabilities





& Operations



Predictive Methods for Carbon Storage



Smart Grid Monitoring





Cybersecurity



Offshore Pipeline and Infrastructure Integrity



Predictive Methods to ID
Rare Earth Elements



Data Discovery, Labeling, Transformation



Prediction of Subsurface Properties and Geohazards



AI/ML Scientific Computing



AI/ML must be applied appropriately (SAMI)





BIG DATA

The Parable of Google Flu: Traps in Big Data Analysis

David Lazer, 1,2* Ryan Kennedy, 1,34 Gary King, 3 Aless andro Vespignani 3,56

T n February 2013, Google Flu Trends (GFT) made headlines ■ but not for a reason that Google executives or the creators of the flu tracking system would have hoped. Nature reported that GFT was predicting more than double the proportion of doctor visits for influenza-like illness (ILI) than the Centers for Disease Control and Prevention (CDC), which bases its estimates on surveillance reports from laboratories across the United States (1, 2). This happened despite the fact that GFT was built to predict CDC reports. Given that GFT is often held up as an exemplary use of big data (3, 4), what lessons can we draw from this error?

The problems we identify are not limited to GFT. Research on whether search or social media can predict x has become common-

place (5-7) and is often put in sharp contrast

surement and construct validity and reli-

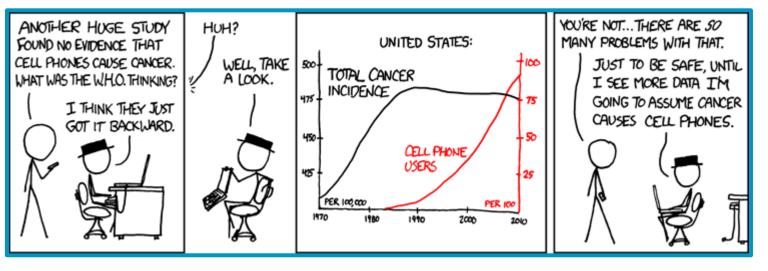
Ever

with traditional methods and hypotheses ability and dependencies among data (12). the com

Lazer et al 2014, the parable of google flu: traps in big data analysis, Science

Large err

- AI/ML methods & technologies must be appropriate for the goal
- Results need to be explainable & validated



- Correlation does not equal causation
 - Just because you have an analysis doesn't mean the results are meaningful
- **Uncertainty** is critical
 - Capture, reduce if possible, represent, utilize, quantify





Accelerating FE R&D with AI and ML



SAMI CROSSCUTTING BENEFITS

COMMUNICATE

Share the latest AI **news** and **breakthroughs**



CREATE

Design critical foundational models, platforms, and standards to be applied to various fields





CATALYZE

Act as a **hub** with **AI/ML experts** to help **SME's** R&D Enhance work force **AI/ML fluency**



CONNECT

Coordination with **partnerships**, **IT**, **communication**, and others to benefit the SAMI community

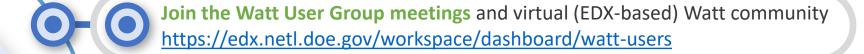


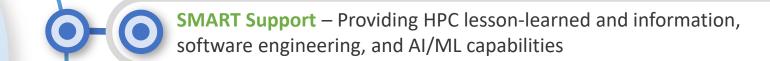
Catalyzing
Opportunities
& Recognition

https://edx.netl.doe.gov/sami



Development of hi-fidelity 3D real-time models for improving prognostics and diagnostics in advanced power system with Cerebras





Workshops and Trainings: AI4CM, NVIDIA bootcamp; Summer Seminar Series "How NETL is using AI/ML to..."

Supporting proposal teams for AITO, DOE SC (BER), EERE, ARPA-e, and others (working with intramural and extramural researchers and partners)

Funding AI/ML IT for (1) Supporting Watt enhancements, (2) AI computing expert to work with R&D teams, (3) Get WATT ready for hosting services (e.g., Jupyter Notebook)



SUCCESSES





HOW CAN YOU CONNECT?

JOINING SAMI IS STRAIGHTFORWARD

Contact Us: SAMI@netl.doe.gov

Learn more about SAMI:

https://edx.netl.doe.gov/sami/

