An IIoT Platform for Agile Development and Deployment of Data-Driven Solutions in E&P*

Tailai Wen¹ and Xinwo Huang¹

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Abstract

In the recent decade, upstream companies' investment in digital transformation and data-driven analytics increased rapidly. Machine learning is expected to bring enhanced insights into E&P workflows that are traditionally based on physics and engineering knowledge. However, the majority of data science projects did not go beyond research papers and few were put in constant operation in actual business processes. Data scientists in upstream companies spend most their time on their local computing environments, which are isolated from real-time data and real users. This gap obstructs continuous testing and improving of data science solutions in real-world scenarios.

Agile methodology is a widely-used product development workflow in Silicon Valley. It emphasizes the importance of continuous feedback from end users and fast cycles of product operationalization. We built an IIoT (Industrial Internet of Things) platform that aims at enabling agile development and deployment of data science solutions in E&P. The platform allows data scientists to operationalize machine learning solutions with minimal IT supports, and it accelerates the feedback-improvement iterations between end users and data scientists effectively.

Core components of this integrated platform includes:

- 1) Edge agent: a software installed on assets (e.g. equipements, sensors, IoT devices) that standardizes real-time data streams. It also hosts models of edge computing which are usually for smart processing of raw data.
- 2) Cloud-based model hub: the center of entire system that connects data, models, and applications. The hub automates enterprise-scale model management (e.g. resourcing, CI/CD, version management). Containerization technology plays a key role here.
- 3) Model deployment software: a tool that enables data scientists to publish models from local environment to model hub and/or to edge agent. Data scientists may operationalize a new solution such that it is ready to end users in just a few minutes.

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4) End-user application: it includes flexible connector to common BI tools, a web-based dashboard builder, and an automated web-based interface renderer.

We tested the platform with our industrial partners in real-world E&P environments. We will highlight a couple of geology-related cases: one on automated well log interpretation, and the other on machine learning assisted seismic imaging.

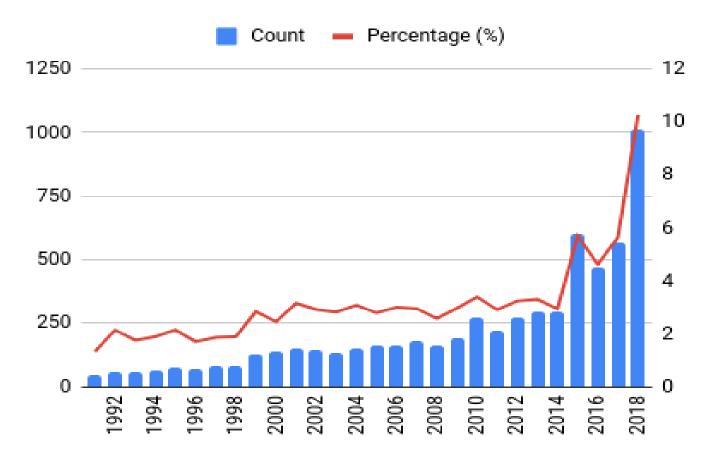
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An IIoT Platform for Agile Development and Deployment of Data-Driven Solutions in E&P

Dr. Tailai Wen*, Dr. Xinwo Huang May 22, 2019 AAPG ACE @ San Antonio, TX

More Attention to Machine Learning / Artificial Intelligence in O&G

Count of papers on machine learning in OnePetro



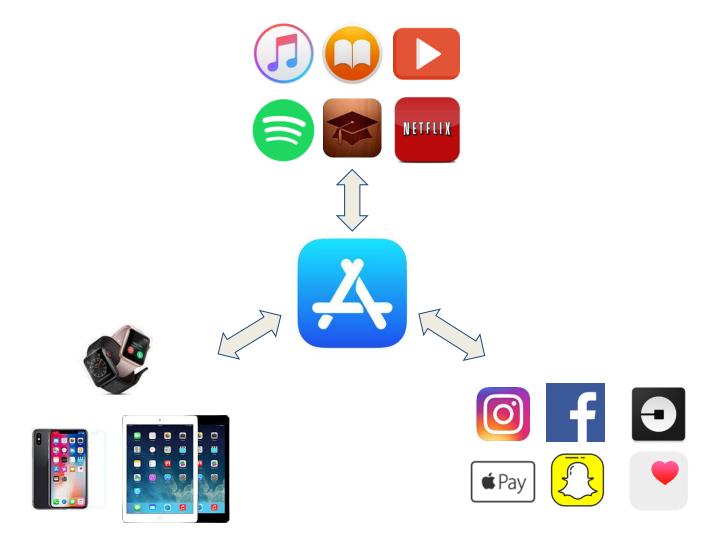
What is Missing?

McKinsey Global Institute interview with 2,135 executives "What is the biggest hurdle for you to achieve digital excellence?"

We are too siloed (i.e., our functions and business units do not work effectively 29 together) Our culture is averse to risk and experimentation Across the organization, we do not have a single, consistent view of what our 25 customers want and need There is no common understanding of our company's culture or strategy across the organization We take a long time to make a decision based on the data we have 20 We are unable to reallocate resources quickly Required sign-offs take a long time to obtain 15 Senior management does not delegate decisions to those in lower levels of the 15 organization Our frontline employees are not empowered to take ownership of decisions in 13 their areas Other Source: McKinsey



Transformation of Ecosystem is Initialized by New Platform



Infrastructure vs. Platform vs. Software

Cloud Clients

Web browser, mobile app, thin client, terminal emulator, ...



Software as a Service (SaaS)

CRM, Email, virtual desktop, communication, games, ...

Platform as a Service (PaaS)

Execution runtime, database, web server, development tools, ...

Infrastructure as a Service (IaaS)

Virtual machines, servers, storage, load balancers, network, ...

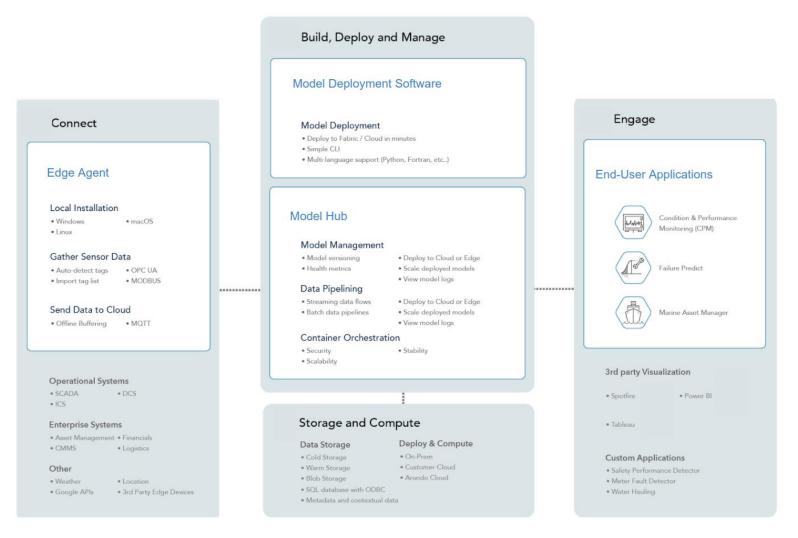
Source: Wikipedia



An Integrated Platform for IIoT

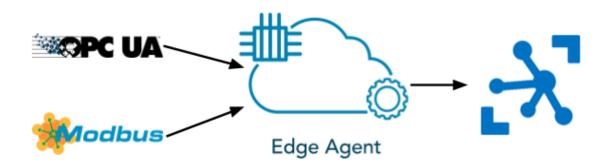
Four key components:

- Cloud-based model management hub
- Edge agent software
- Model deployment software
- End-user interface





Edge Agent for Data Streaming



Industrial Connectors	Edge Agent	Microsoft IOT Hub
 → Supports up to 10 hz frequency from today's main control system interfaces → Easily browse for available tags or upload large csv files with tag information → Build a compute chain to modify or scale any raw data → Timestamping can be handled by the control system or edge agent 	 → Buffer offline data for days (configurable) and push to the cloud when internet connectivity is established → Report health of the system so you can monitor it remotely → Build compute chains to add virtual sensors to your stream → Use a simple viewer to check current streaming data 	 → Out of the box support for Microsoft's IOT Hub → Stream to your own IOT Hub or Arundo's → Cloud connection can be configured in less than a minute → Ability to configure "at least once" delivery confirmation → In poor bandwidth situations data sample rates can be changed

Model Deployment and Management

- Model hub connects data, models, and applications
- Each model is a SaaS deployed at a docker container
- Model deployment software is an interface for data scientists to publish their models as an API (application programming interface)
- Model hub is fully responsible for enterprise-level management work (resourcing, CI/CD, version controls, scheduling, performance monitoring, etc.)



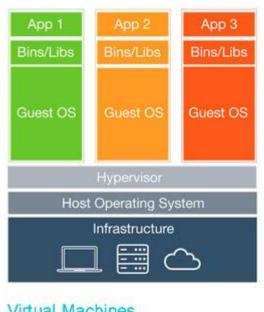
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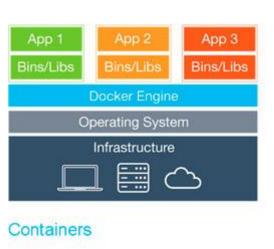
Containerization

- Operating-system-level virtualization technology that has emerged in recent year
- Reduced resource usage, faster, cheaper and more scalable
- Isolated environments, more secure

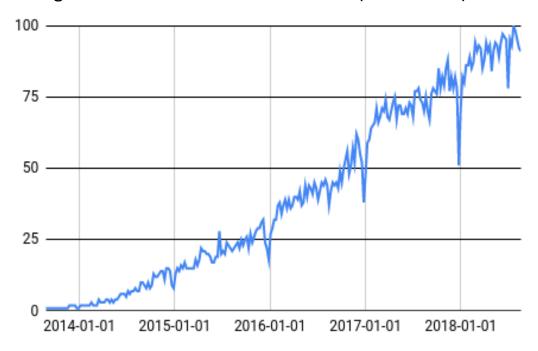




Source: Docker.com



Google searches of "docker container" (normalized)





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Model Deployment

A Python library to wrap a machine learning model into a deployment-ready model

```
dendpoint()
dergument("demoSensor_01", type=float)
dergument("demoSensor_02", type=float)
def equipment_status_flag", type=float)
float
flo
```

• A CLI (command line interface) to push models into model hub

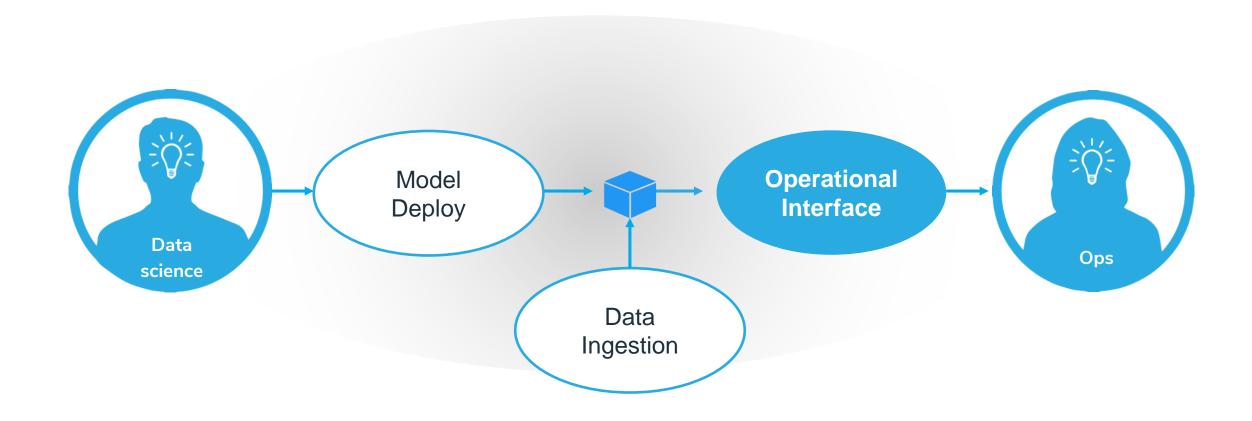
```
(arundo) tailaiw@tailaiw-XPS-15-9560:~/work/arundo-demo/end-to-end/equipment_status_predictor$ composer publish
Validating model...
Model validated!
Bundling model...
Bundling app in folder /home/tailaiw/work/arundo-demo/end-to-end/equipment_status_predictor
Making zip file: equipment_status_predictor.zip
Model bundled!
Creating model doc...
Model doc created id = 813999fd-574d-a971-8d29-3a549d1bb4ee
Uploading model zip file...
Model upload succeeded
Build triggered
```

Model Deployment and Management

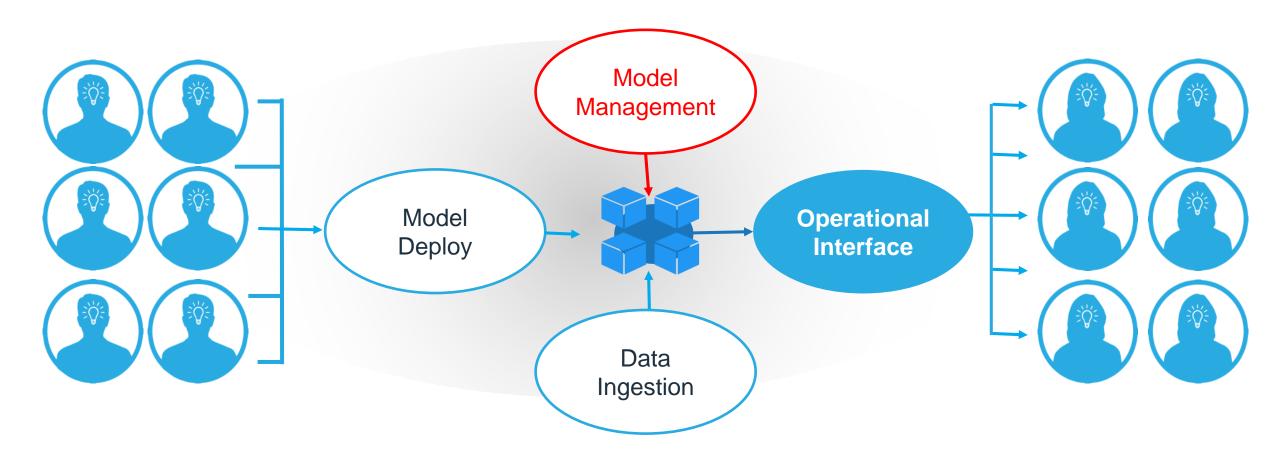
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Model Management in Individual Scale



Model Management in Enterprise Scale



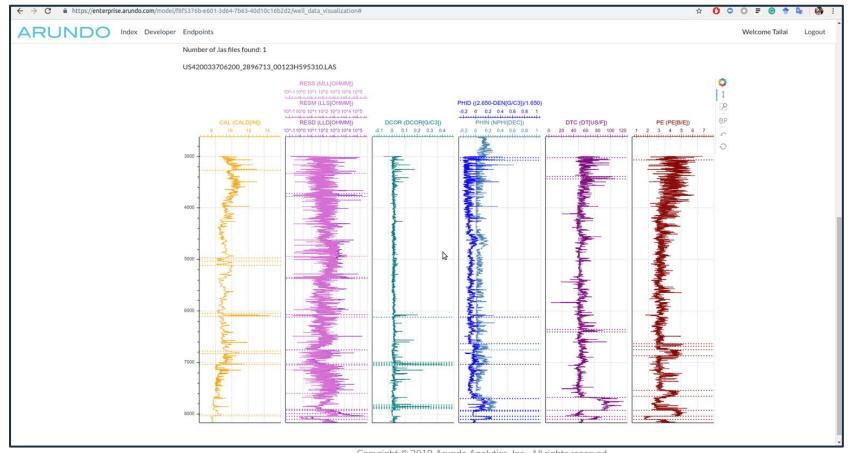
End-User Interface

- Flexibility to connect common BI tools
- Customized web-based dashboard
- Customized user interface



End-User Interface

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Why is This Important to Data Scientists in E&P Companies?

Common skill set of a data scientist in 2019

Includes:

- Python, R, MATLAB, Julia, ...
- Jupyter notebook, PyCharm, ...
- Machine learning, deep learning, statistics, ...
- Scikit-learn, Keras, TensorFlow, PyTorch,

Does NOT include:

- Docker, Kubernetes
- OPC, Modbus, MQTT, Kafka, ...
- NodeJS, PHP, C#, GoLang, ...
- Authentication, network security
- Travis CI, Jenkins, Circle CI, ...
- ReactJS, AngularJS, VueJS, CSS, ...



Why is This Important to E&P Companies?

Building an agile ecosystem of data-driven decision making



Loop needs to be closed

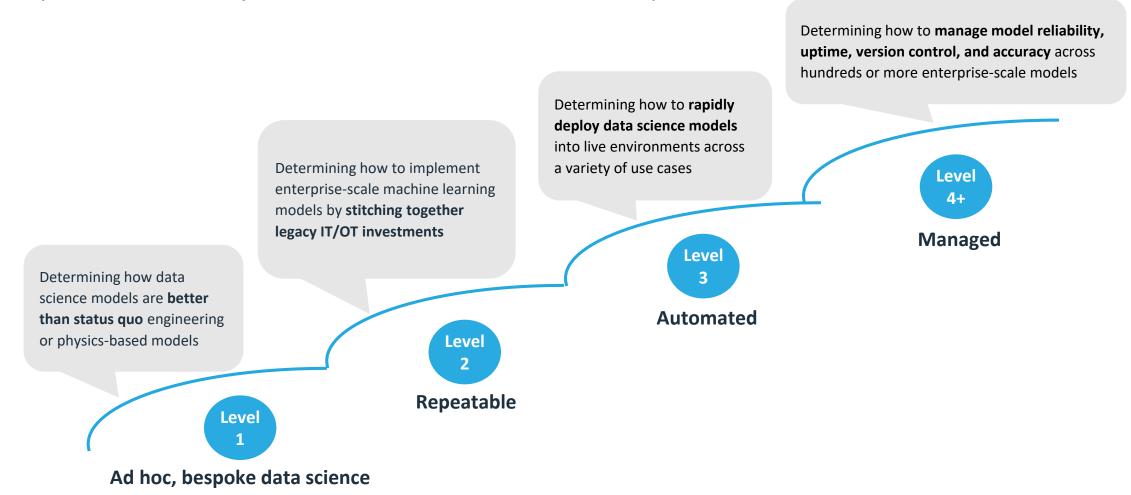
Operationalization needs to be accelerated

MVP (minimum viable product) design needs to replace traditional R&D mindset



Why is This Important to E&P Companies?

Operational maturity curve for industrial data science adoption

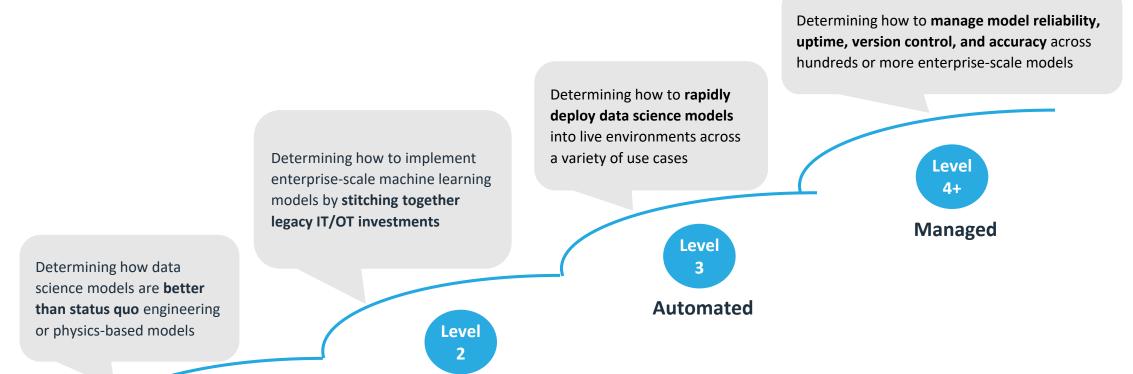


Why is This Important to E&P Companies?

Level

Ad hoc, bespoke data science

Operational maturity curve for industrial data science adoption



You don't need to go through step by step!

"A common mistake is thinking that legacy IT systems have to be integrated first."

---Ten Red Flags Signaling Your Analytics Program Will Fail by McKinsey



Repeatable

Why is This Important to E&P Sector?

