



Machine Learning on AWS - Technical

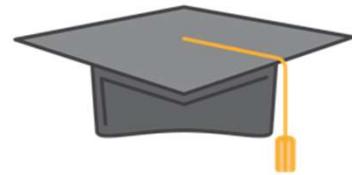
AWS Solutions Training for Partners

Vijay
AWS Partner Trainer

Machine Learning



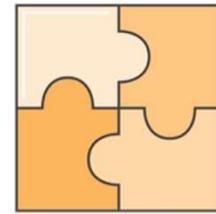
Prediction is the process of filling in missing information; it uses data you have to generate data you don't have.



Learning



Language



Perception



Problem
Solving



Insight

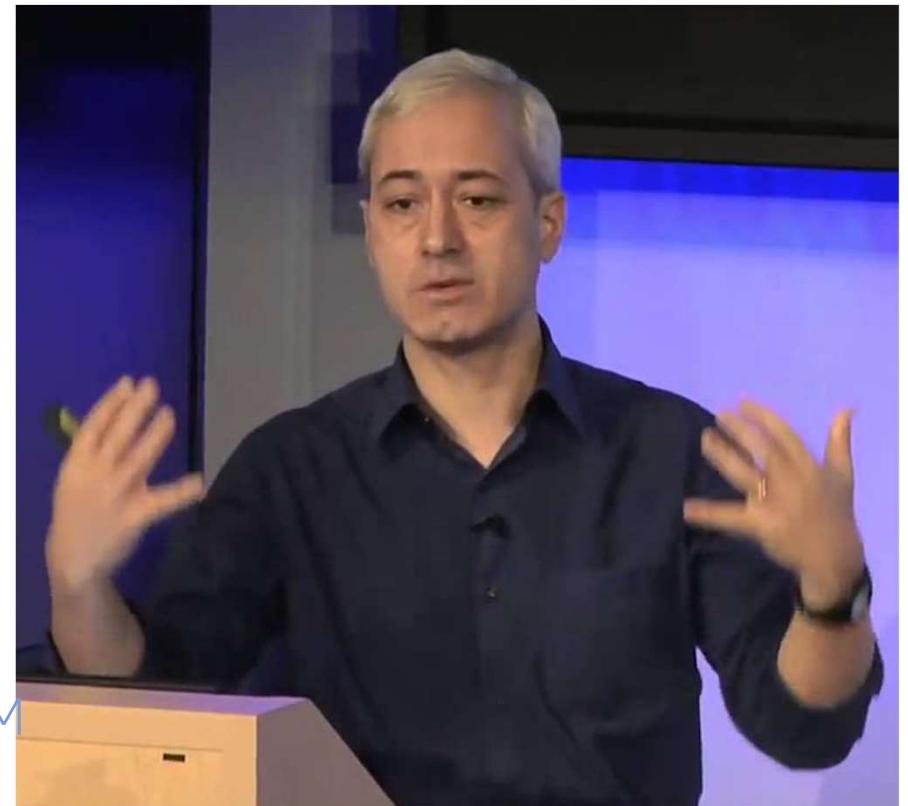
The Artificial Intelligence Landscape

Five Tribes / Two Breakthroughs



Tribe	Origins	Algorithm
Bayesians	Statistics	Probabilistic
Analogizers	Psychology	Kernel
Symbolists	Logic	Inverse Deduction
Evolutionaries	Biology	Genetic
Connectionists	Neuroscience	Back Propagation

- Computer Vision : [CNNs](#)
 - Static / Unstructured
- Natural Language Processing : [RNNs / LSTM](#)
 - Sequential / Structured





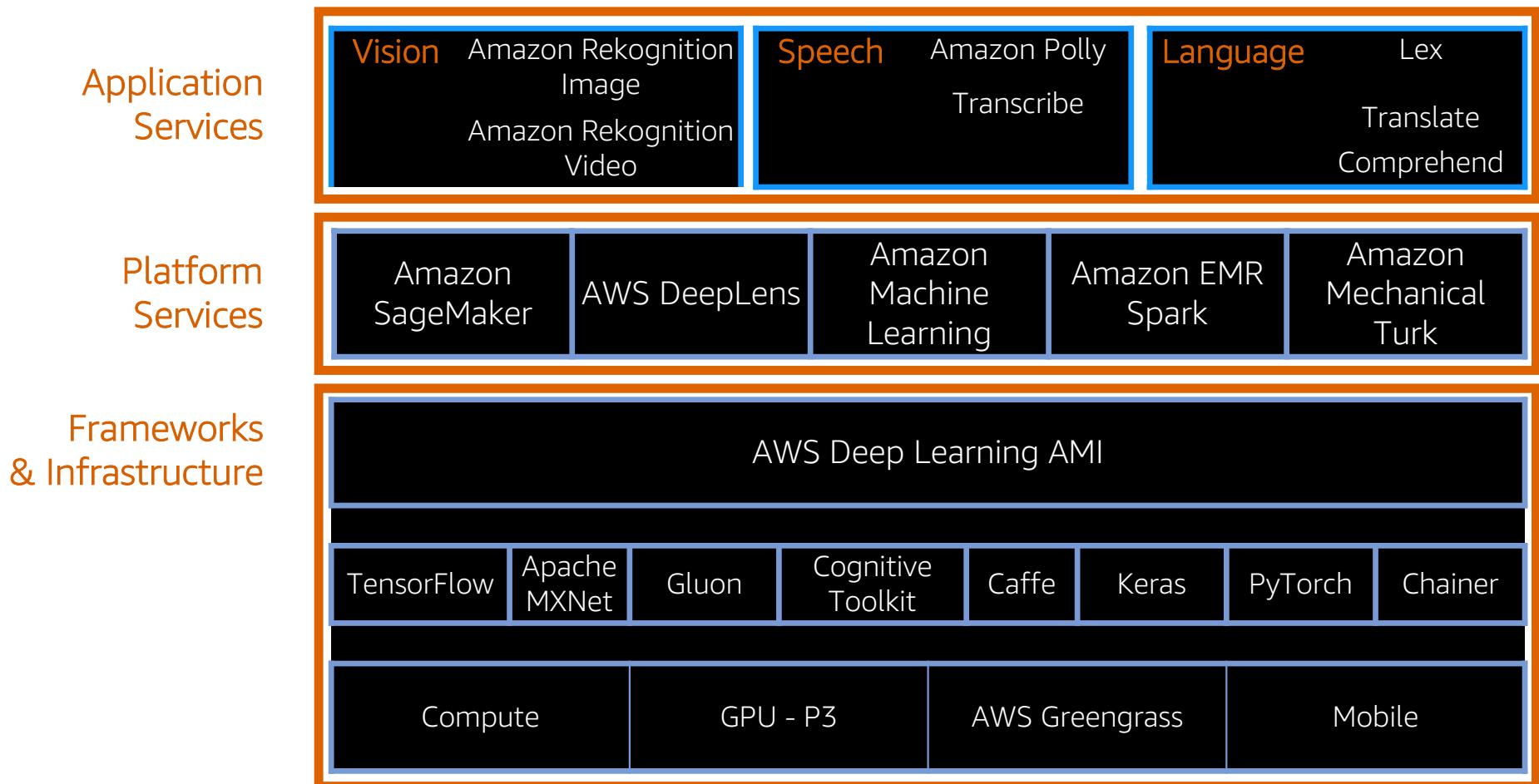
AWS Mission

Put machine learning in the hands of every developer, data scientist and architect

Customers Running ML on AWS



The AWS Machine Learning Stack



The AWS Machine Learning Stack



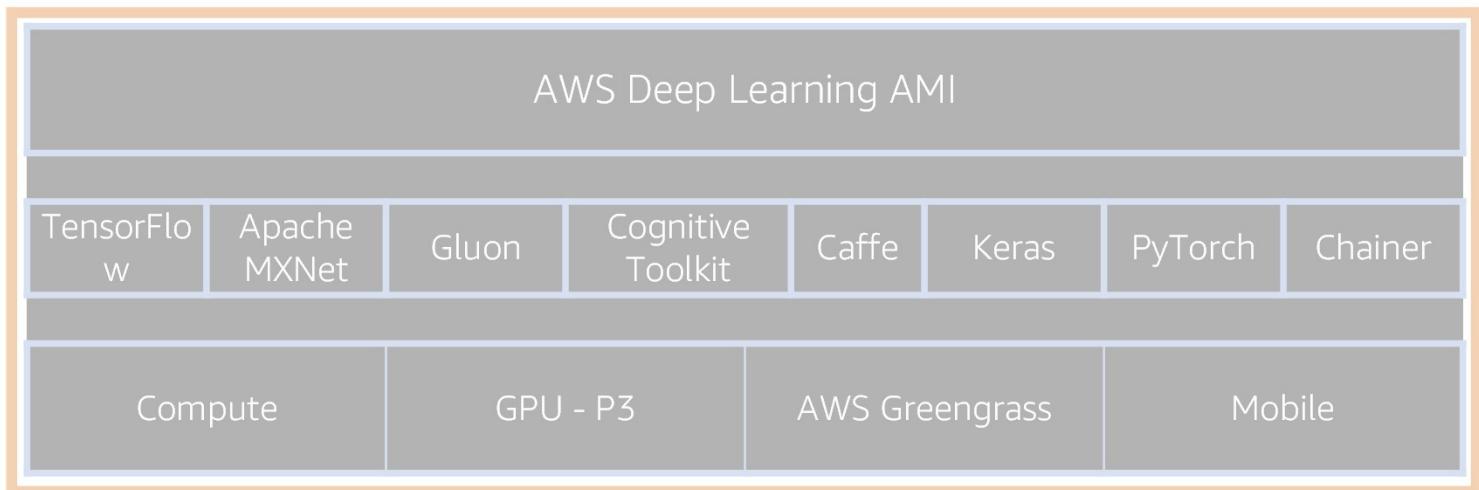
Application Services



Platform Services



Frameworks & Infrastructure





Demo 1: Vision Services

The AWS Machine Learning Stack



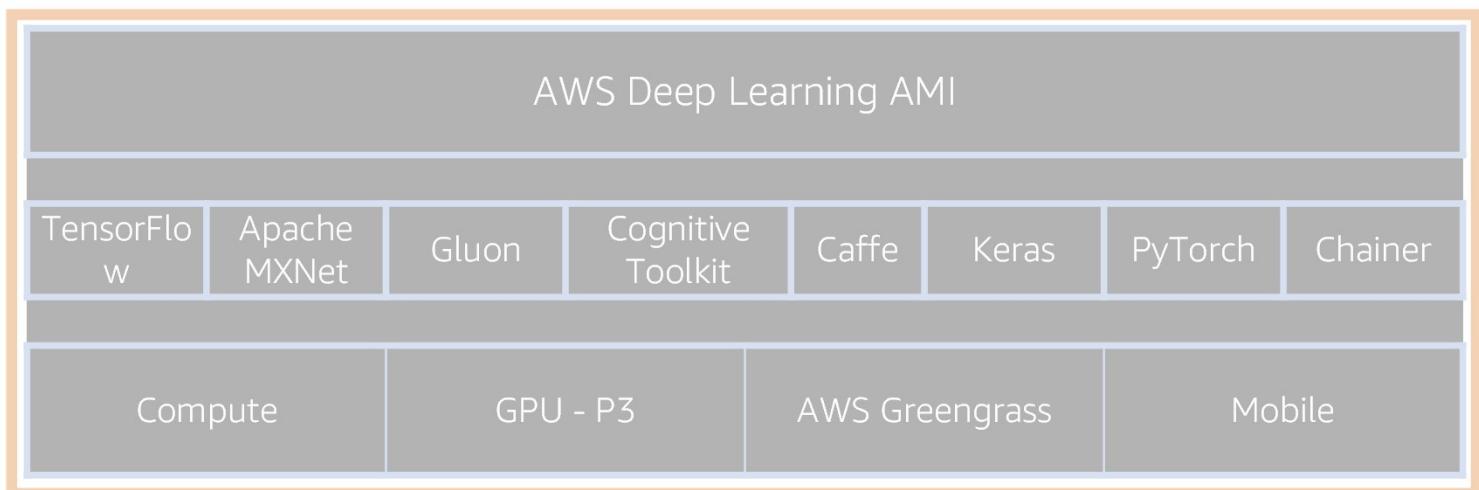
Application Services



Platform Services



Frameworks & Infrastructure





Demo 2: Language Services

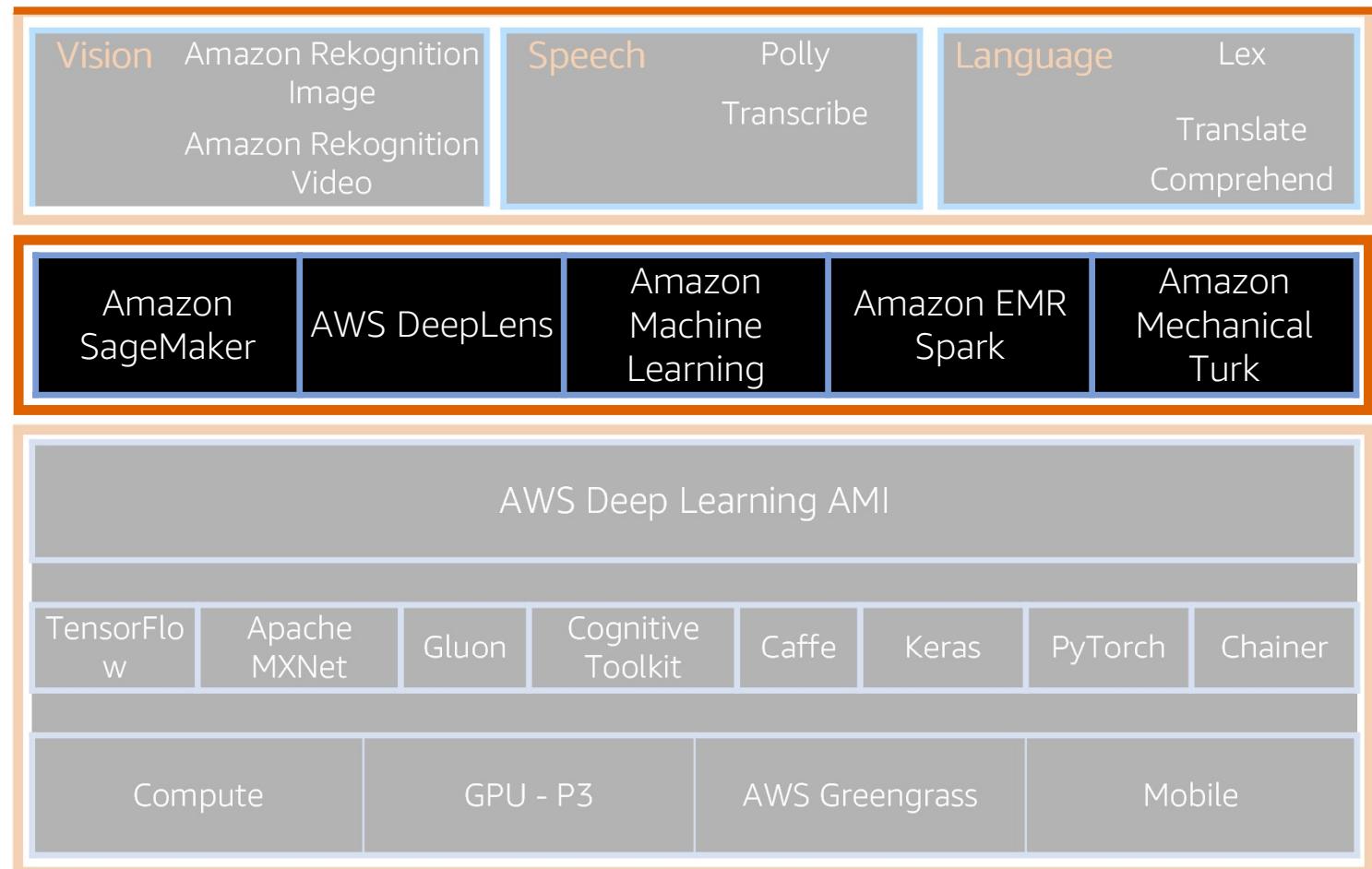
The AWS Machine Learning Stack



Application Services

Platform Services

Frameworks & Infrastructure



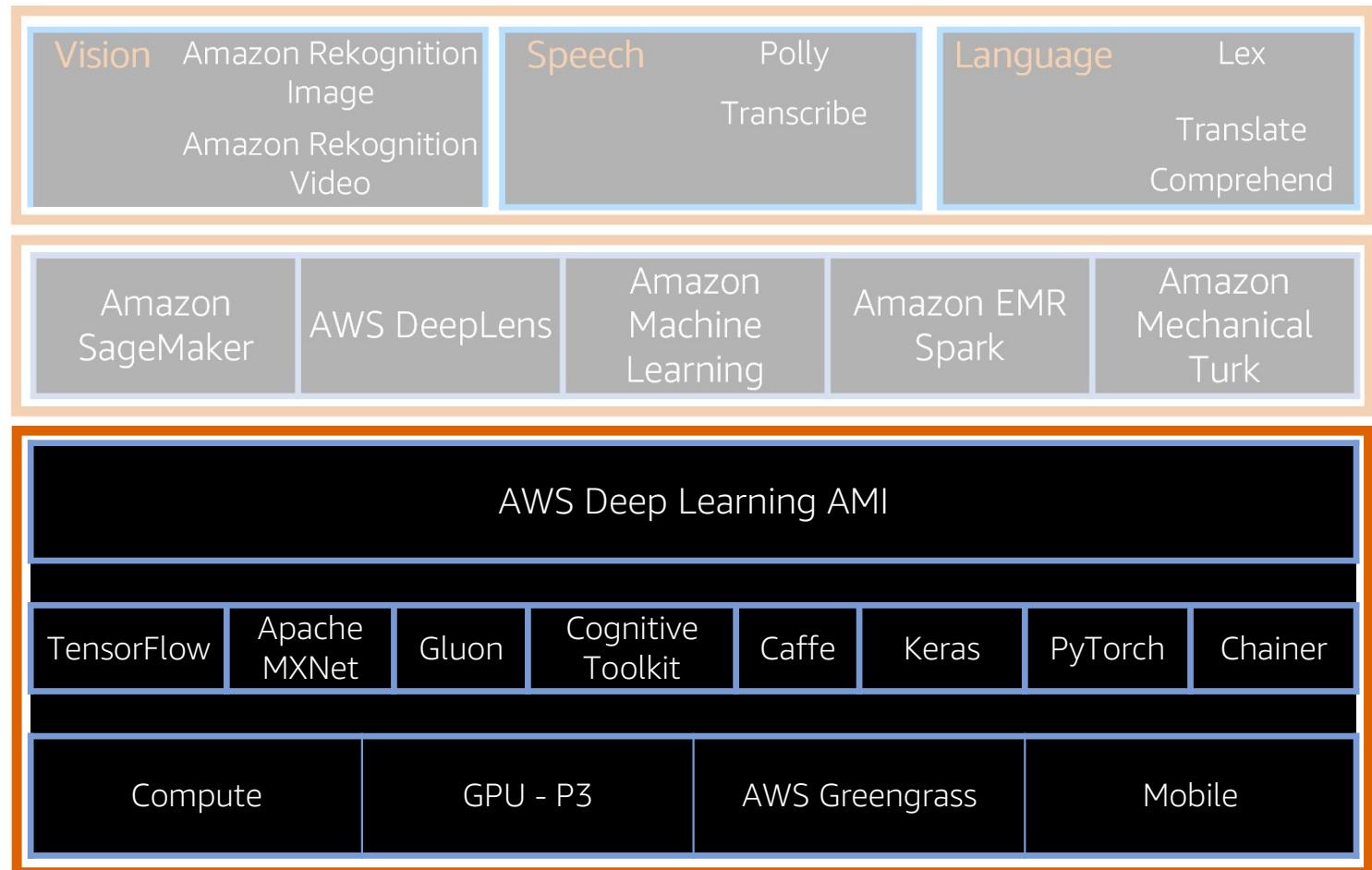
The AWS Machine Learning Stack



Application Services

Platform Services

Frameworks & Infrastructure





Demo 3: Deep Learning AMI

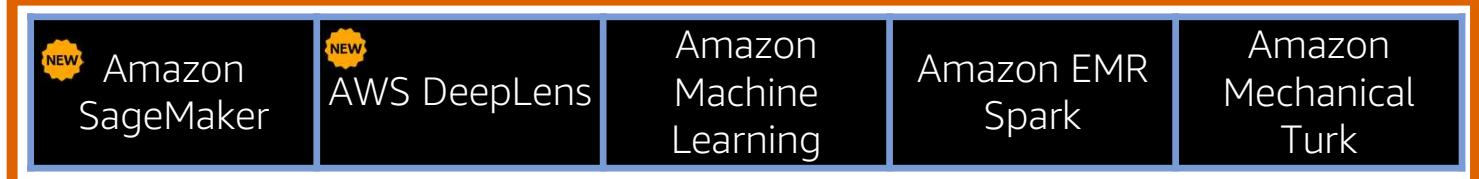
The AWS Machine Learning Stack



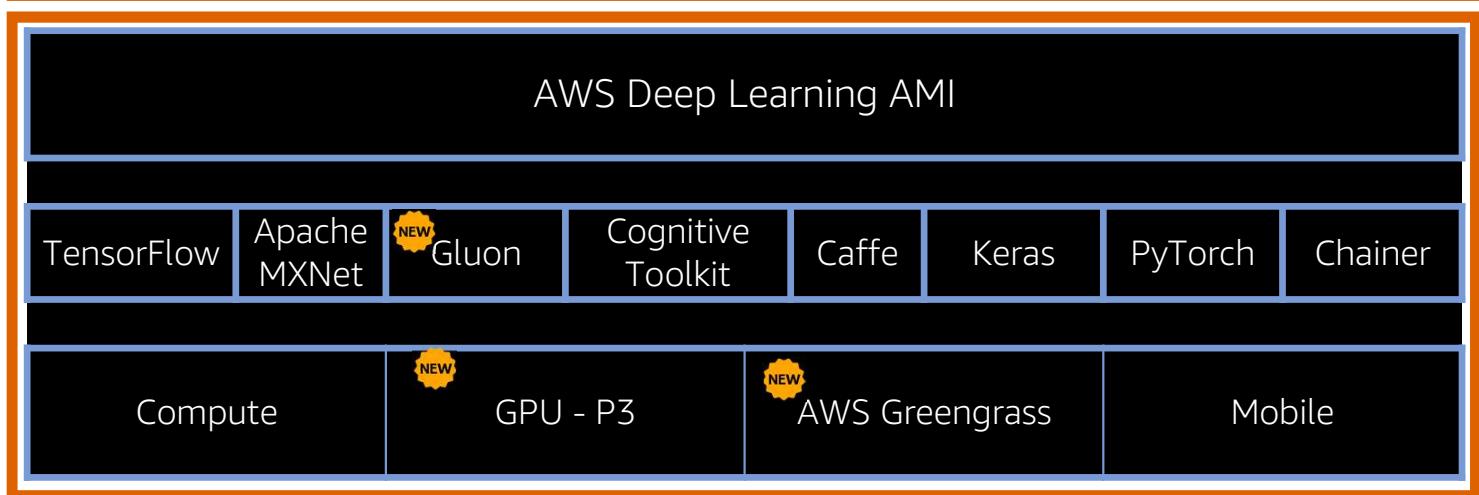
Application Services



Platform Services



Frameworks & Infrastructure



Amazon EC2 P3 Instances



The fastest, most powerful GPU instances in the cloud

- Up to 8 NVIDIA Tesla V100 GPUs
 - 16GB GPU memory with 900 GB/sec peak bandwidth
- 1 PetaFLOPs of computational performance
 - 14x better than P2
- 300 GB/s GPU-to-GPU communication (NVLink)
 - 9X better than P2

Airbnb

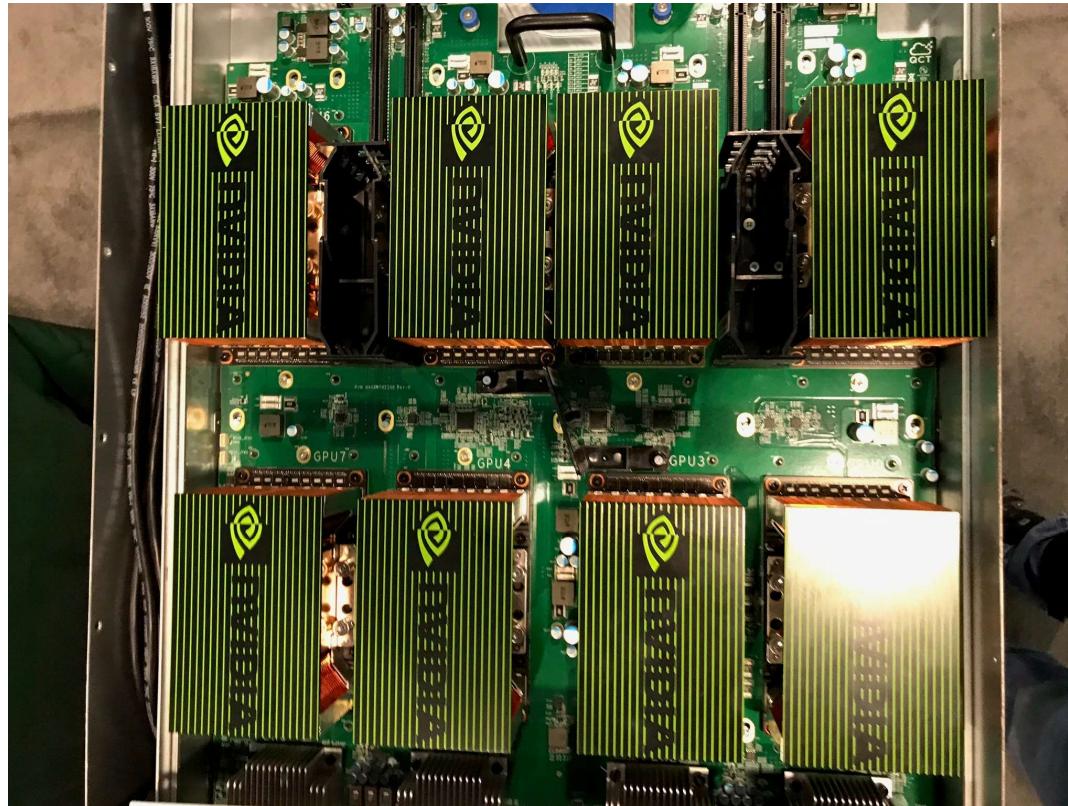
Toyota Research Institute

OpenAI

Amazon EC2 P3 Instances



Amazon EC2 P3 Instances



AWS Deep Learning Amazon Machine Image (AMI)



- Get started quickly with easy-to-launch tutorials
- Hassle-free setup and configuration
- Pay only for what you use – no additional charge for the AMI
- Accelerate your model training and deployment
- Support for popular deep learning frameworks

TensorFlow, MXNet, Gluon, Keras, Caffe2, PyTorch, Zendesk, Matric Analytics, SCDM, etc.

Amazon ML Solutions Lab



Lots of companies doing
Machine Learning



Lack ML
expertise



Unable to unlock business
potential

Amazon ML Solutions Lab



Lots of companies doing
Machine Learning



Lack ML
expertise



Unable to unlock business
potential

Amazon ML Solutions
Lab provides the
missing ML expertise



Brainstorming



Modeling



Education

Amazon ML Lab Customers



Johnson & Johnson

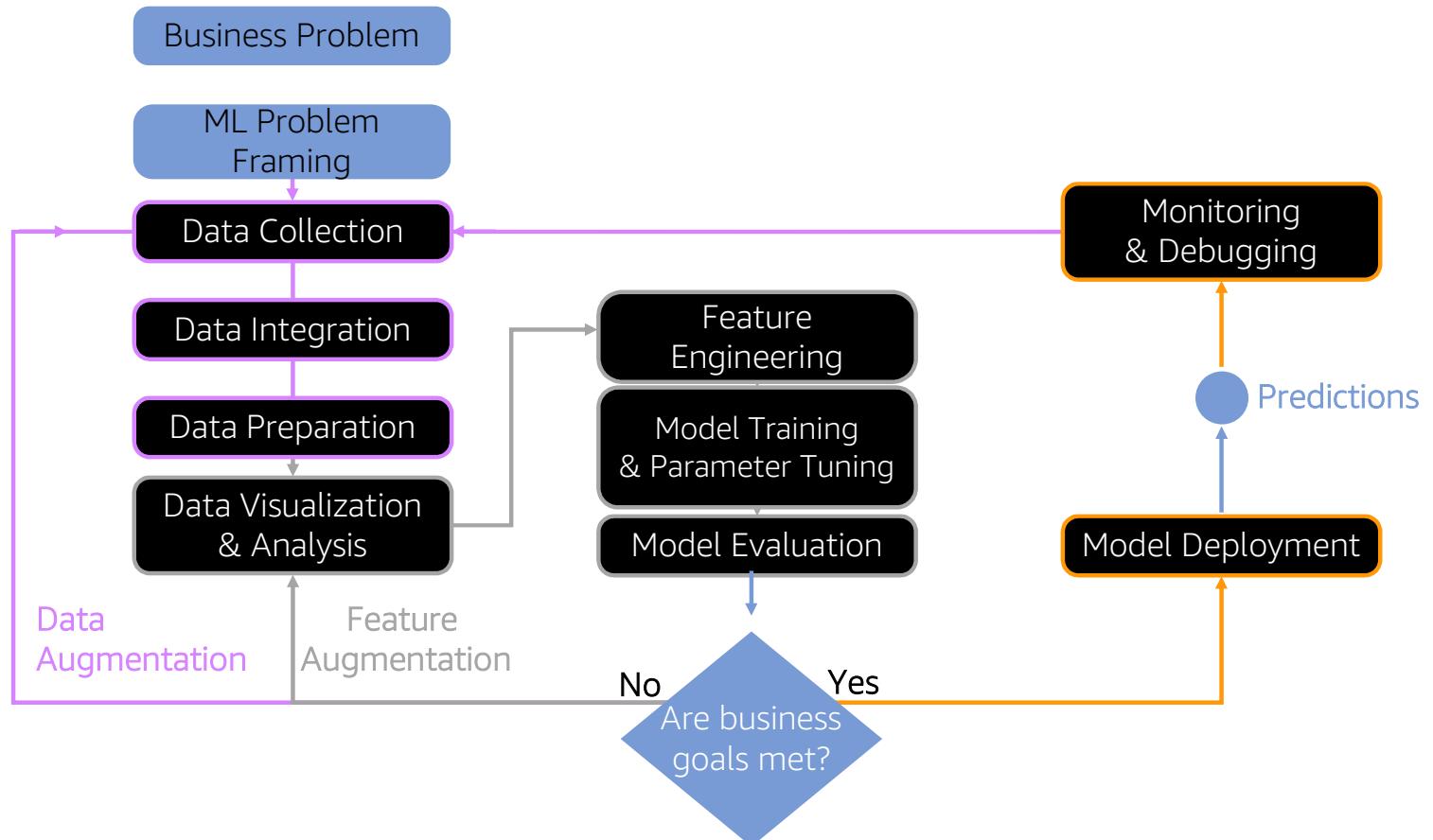
Toyota Research Institute

Washington Post



The Machine Learning Process

The Machine Learning Process



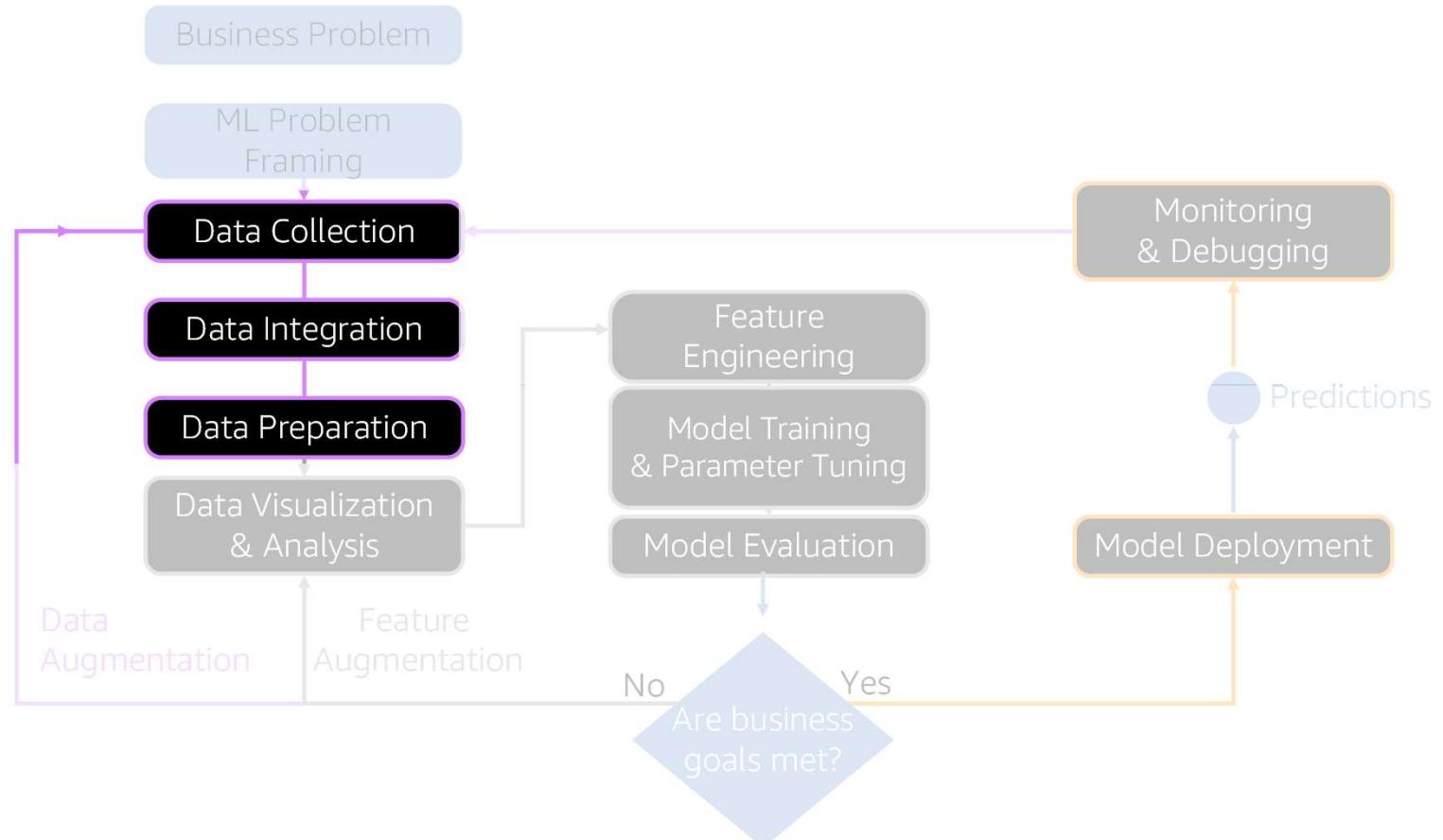
The ML Process

Integration: The Data Architecture



Build the data platform:

- Amazon Simple Storage Service (Amazon S3)
- Amazon Athena
- Amazon EMR
- Amazon Redshift
- AWS Glue

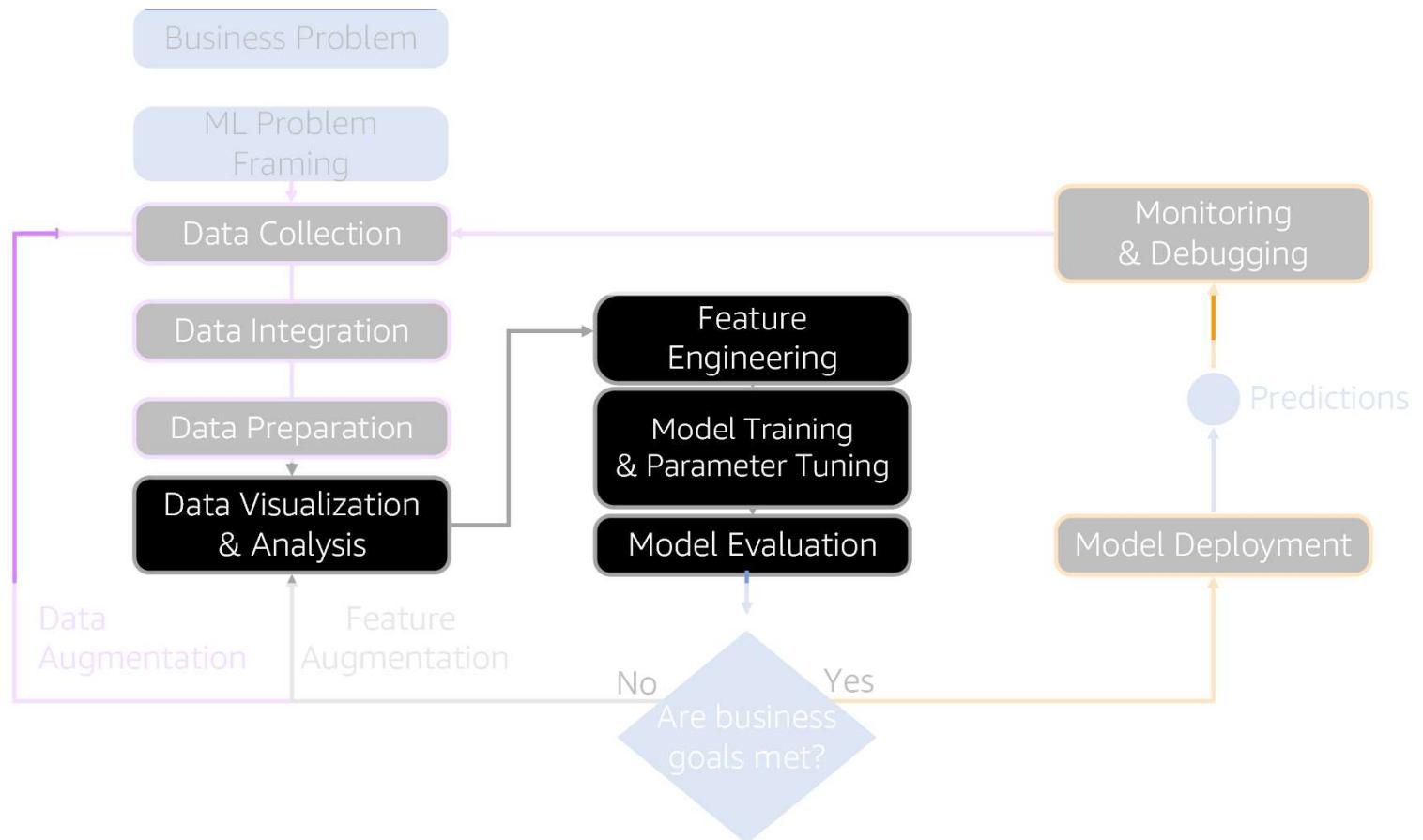


The ML Process

The Model Training: Undifferentiated Heavy Lifting



- Setup and Manage
 - Notebook Environments
 - Training Clusters
- Write Data Connectors
- Scale ML algorithms to large datasets
- Distribute ML training algorithm to multiple machines
- Secure model artifacts

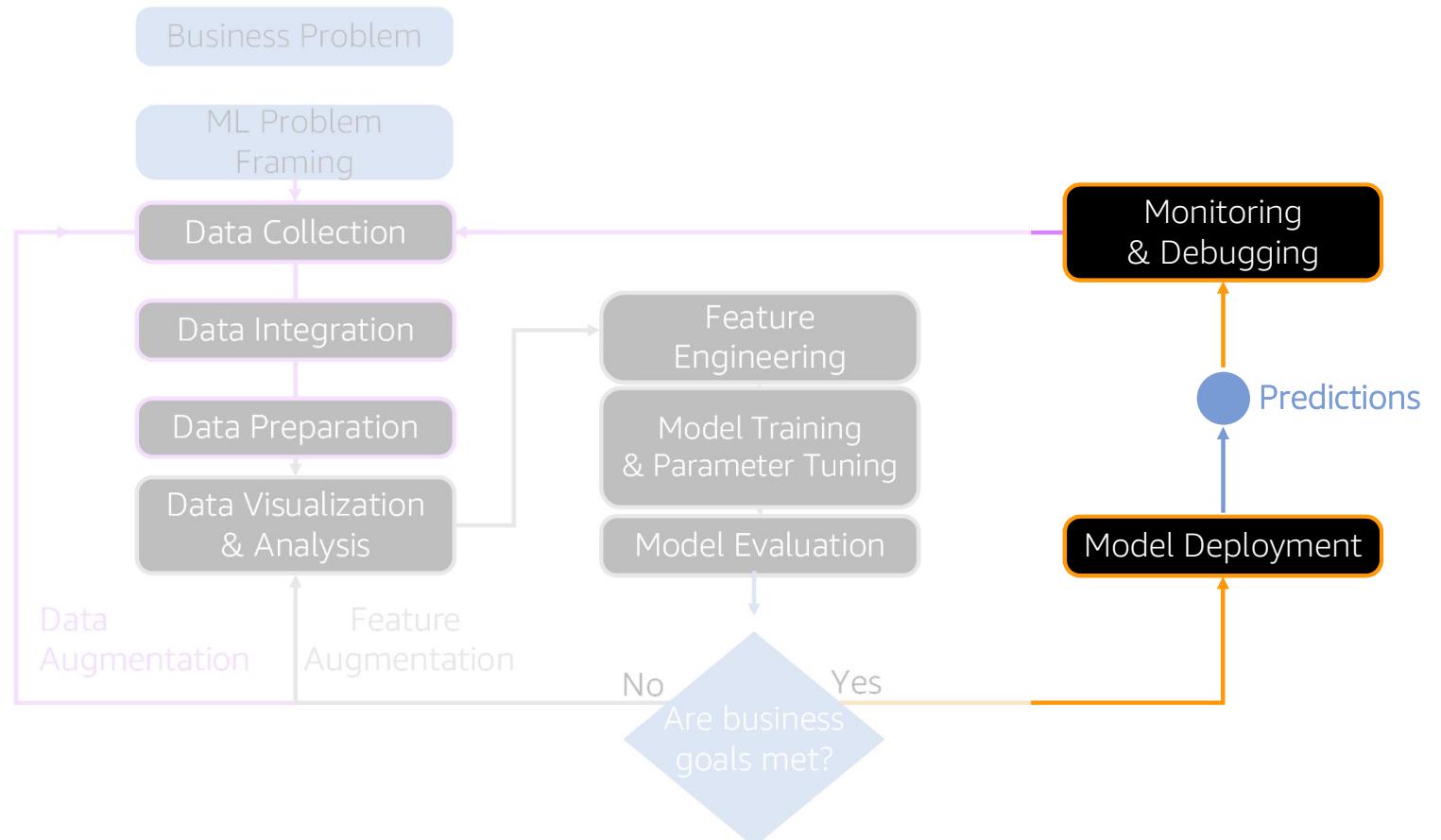


The ML Process

DevOps: Undifferentiated Heavy Lifting



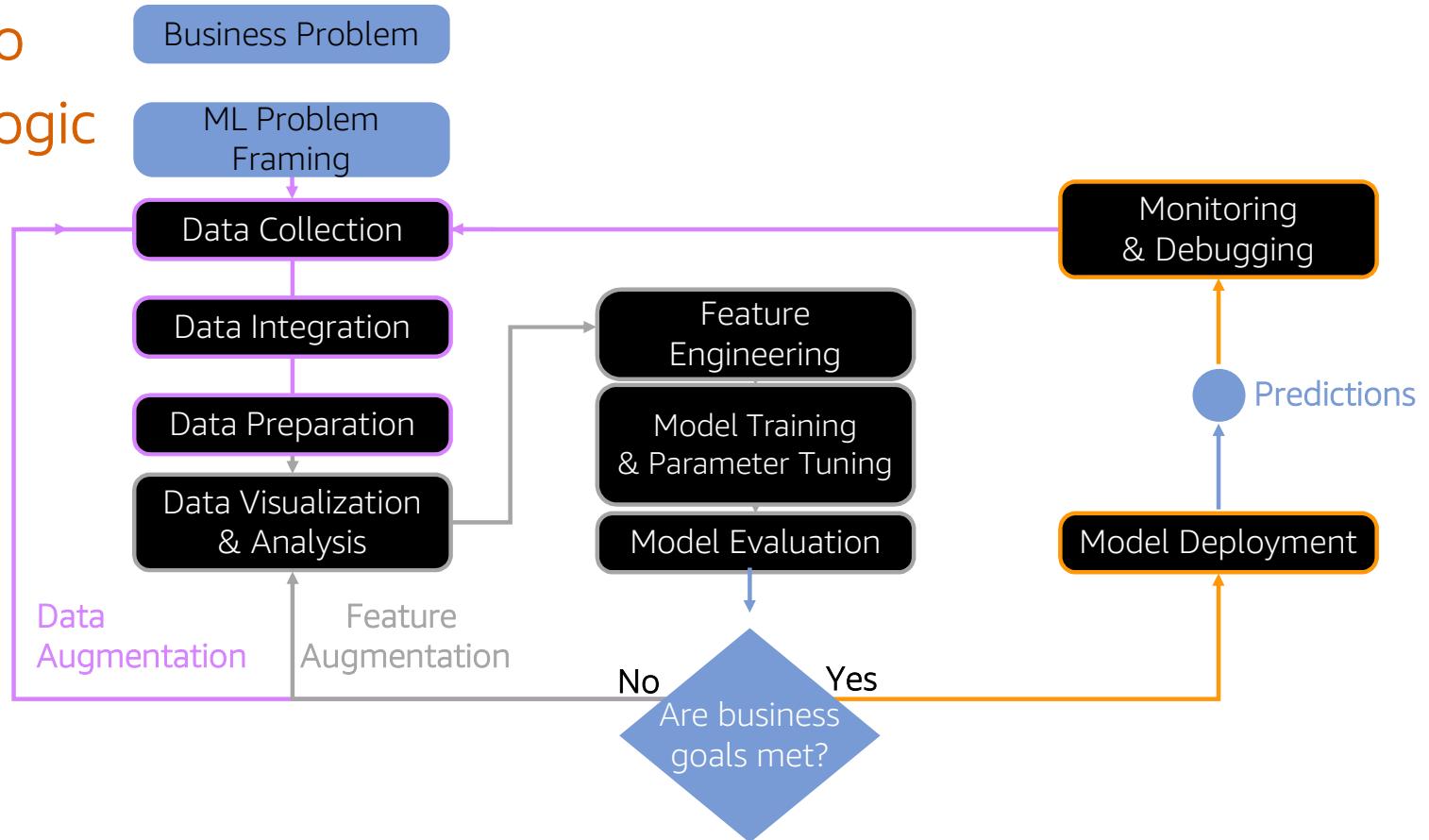
- Setup and Manage Inference Clusters
- Manage and Scale Model Inference APIs
- Monitor and Debug Model Predictions
- Models versioning and performance tracking
- Automate New Model version promotion to production (A/B testing)



Why Amazon SageMaker?



You Only Have to
Write Business Logic





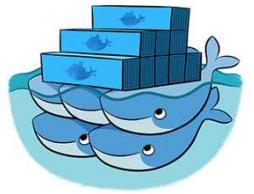
Amazon SageMaker

A Fully-Dockerized Lifecycle

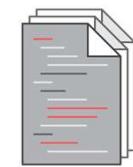
From Discovery to Development and Deployment



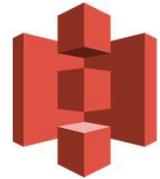
Data Scientists



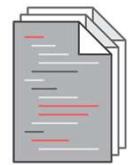
Amazon SageMaker



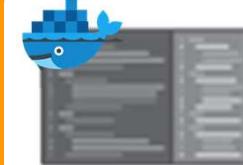
Model Artifacts



Amazon S3



Training Data



Training Algorithm



Inference Engine



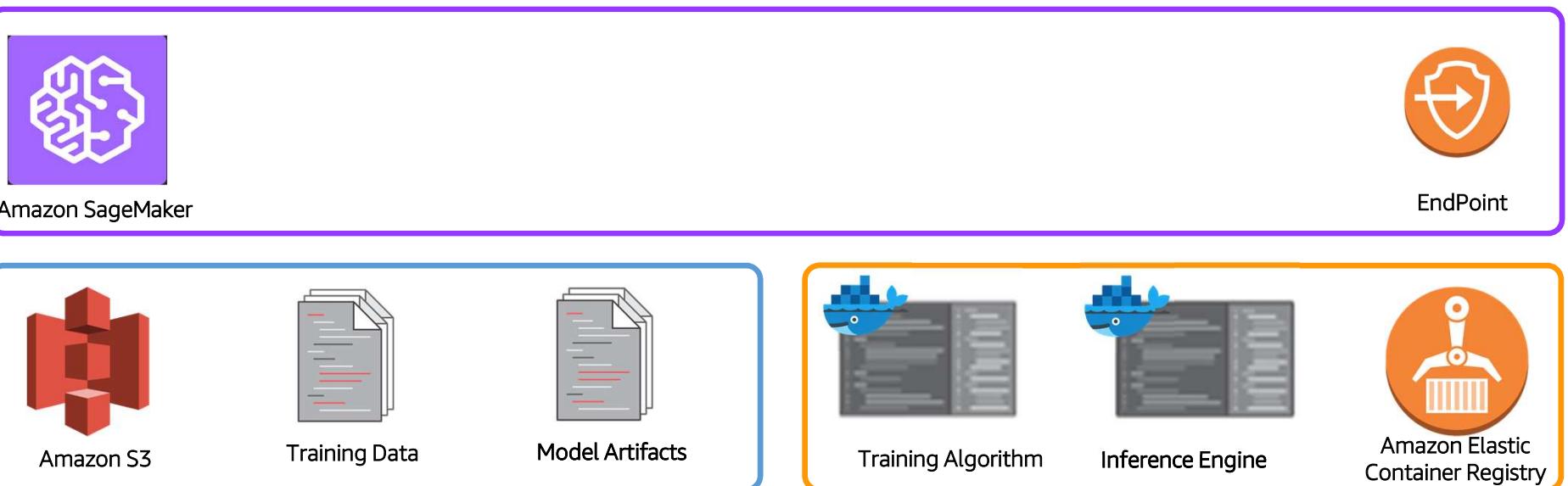
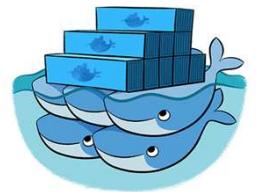
Amazon Elastic
Container
Registry

A Fully-Dockerized Lifecycle

From discovery to development and deployment



Developers and Operations



A Fully-Dockerized Lifecycle

From discovery to development and deployment



Delighted Customers



Predictive Model

Amazon SageMaker

Launch Customers



Intuit

Digital Globe

ZipRecruiter

Hotels.com

Thomson Reuters

Customer Example: Intuit



INTUIT

"With Amazon SageMaker, we can accelerate our Artificial Intelligence initiatives at scale by building and deploying our algorithms on the platform. We will create novel large-scale machine learning and AI algorithms and deploy them on this platform to solve complex problems that can power prosperity for our customers."

Ashok Srivastava, Chief Data Officer, Intuit

Key Benefits of Amazon SageMaker at Intuit



From

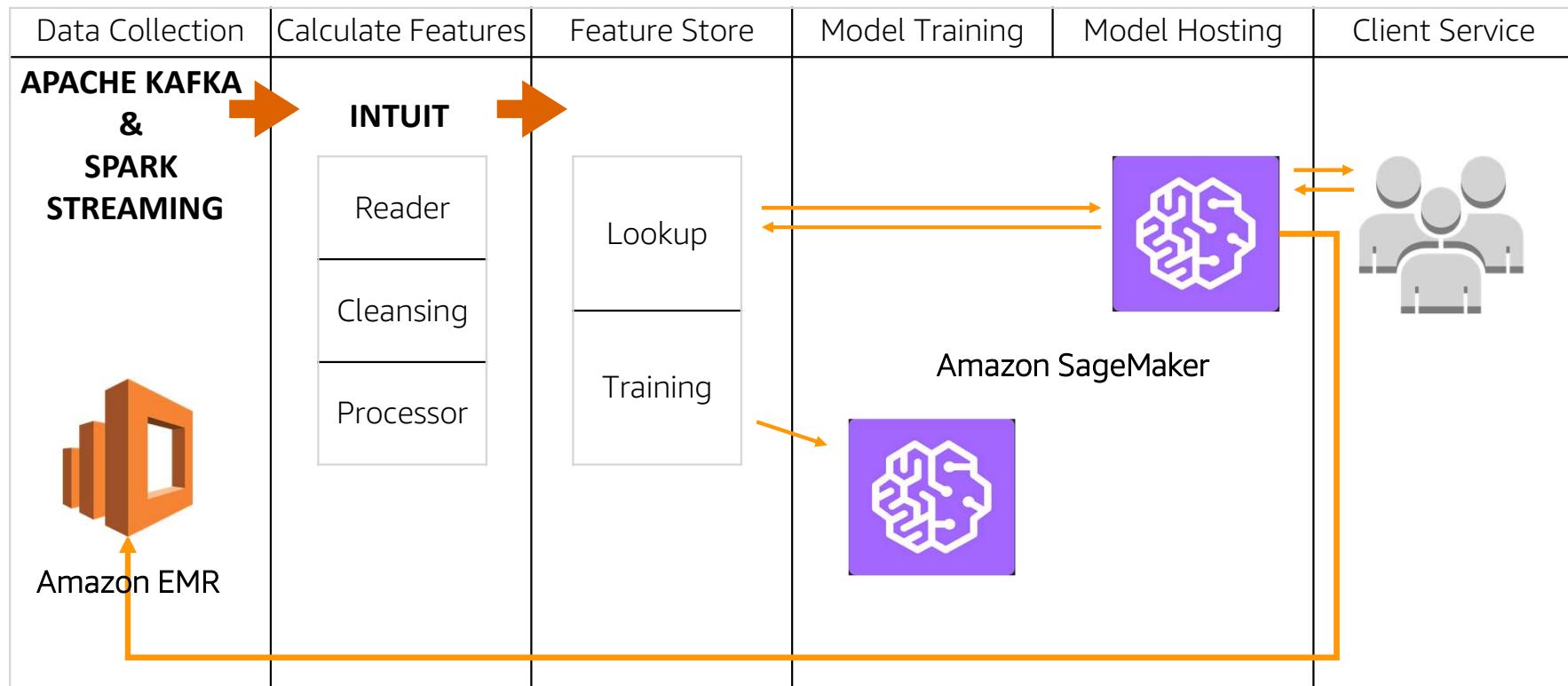
To

Ad-hoc setup and management of notebook environments

Limited choices for model deployment

Competing for compute resources across teams

Fraud Detection using SageMaker



Customer Example: DigitalGlobe



DigitalGlobe

"As the world's leading provider of high-resolution Earth imagery, data and analysis, DigitalGlobe works with enormous amounts of data every day. DigitalGlobe is making it easier for people to find, access, and run compute against our entire 100PB image library, which is stored in AWS's cloud, to apply deep learning to satellite imagery. We plan to use Amazon SageMaker to train models against petabytes of Earth observation imagery datasets using hosted Jupyter notebooks, so DigitalGlobe's Geospatial Big Data Platform (GBDX) users can just push a button, create a model, and deploy it all within one scalable distributed environment at scale."

Dr. Walter Scott, CTO of Maxar Technologies and founder of DigitalGlobe

Customer Example: ZipRecruiter



ZipRecruiter

"We're focused on making it faster and easier than ever to hire and get hired, training our machine learning algorithms against hundreds of millions of historical transactional activities in order to deliver highly relevant job matches as quickly as possible. Amazon SageMaker provided us with an answer to problems we had with ML workflow management, allowing us to train, evaluate and deploy models in a flexible way. In addition, Amazon SageMaker's modularity provides the ability to build and create models independently, which is a compelling feature for ZipRecruiter."

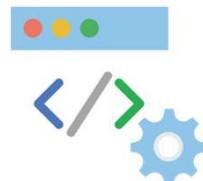
Avi Golan, VP of Engineering, ZipRecruiter

Amazon SageMaker's Components



Amazon SageMaker

1



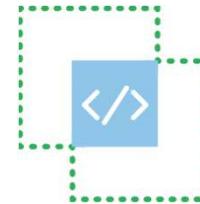
Notebook Instances

2



Algorithms

3



ML Training Service

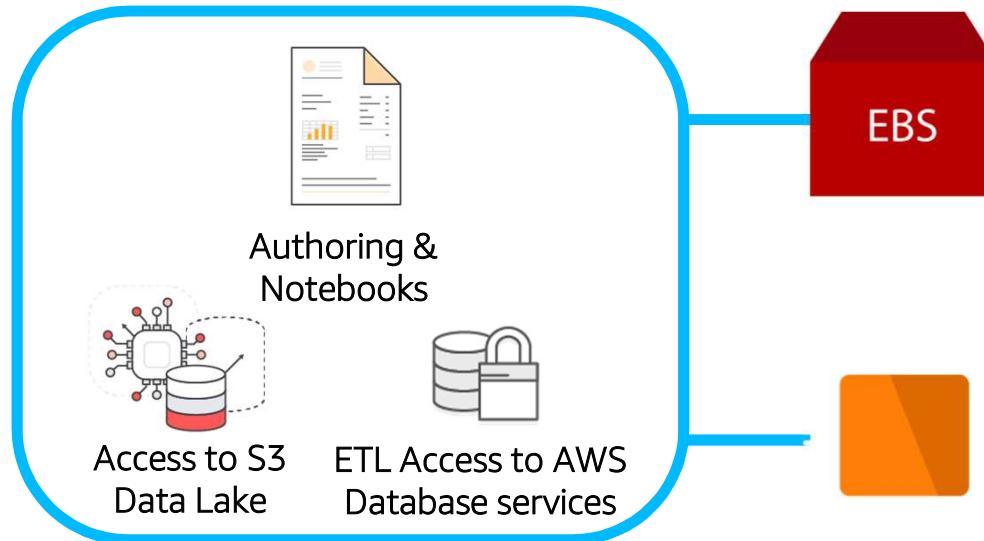
4



ML Hosting Service

SageMaker Notebook Instances

Zero Setup for Exploratory Data Analysis



Just add data!

- Recommendations/Personalization
- Fraud Detection
- Forecasting
- Image Classification
- Churn Prediction
- Marketing Email/Campaign Targeting
- Log processing and anomaly detection
- Speech to Text
- More...



Demo 4: A simple Jupyter Notebook

Pythagorean Theorem



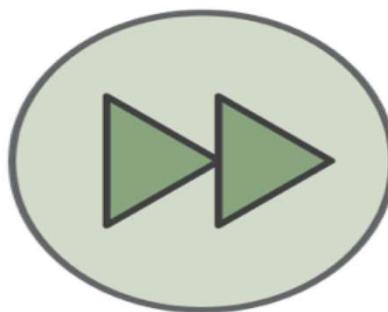
Demo 5: Predicting AWS Spot Pricing

SageMaker Built-in Algorithms

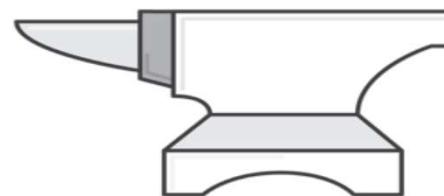
10x Faster



Streaming datasets, for
cheaper training



Train faster, in a single
pass



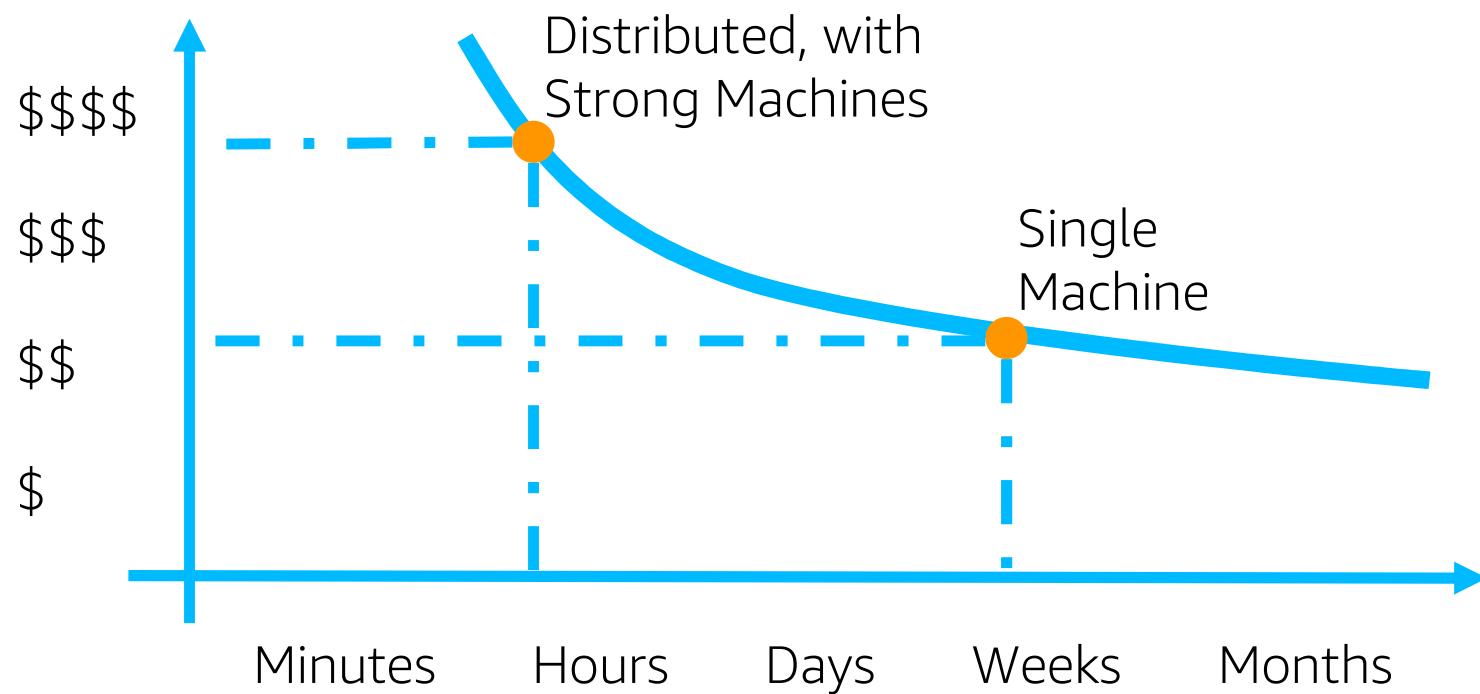
Greater reliability on
extremely large
datasets



Choice of several ML
algorithms

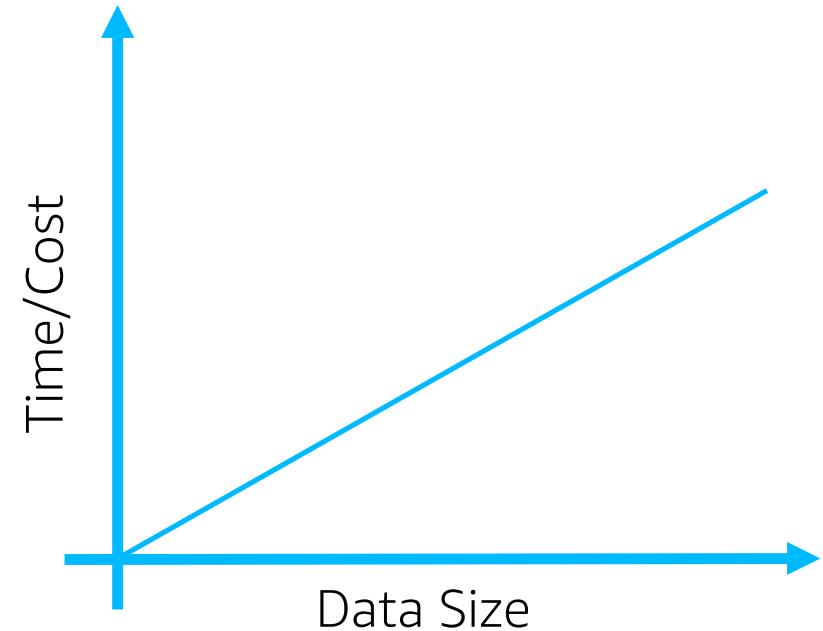
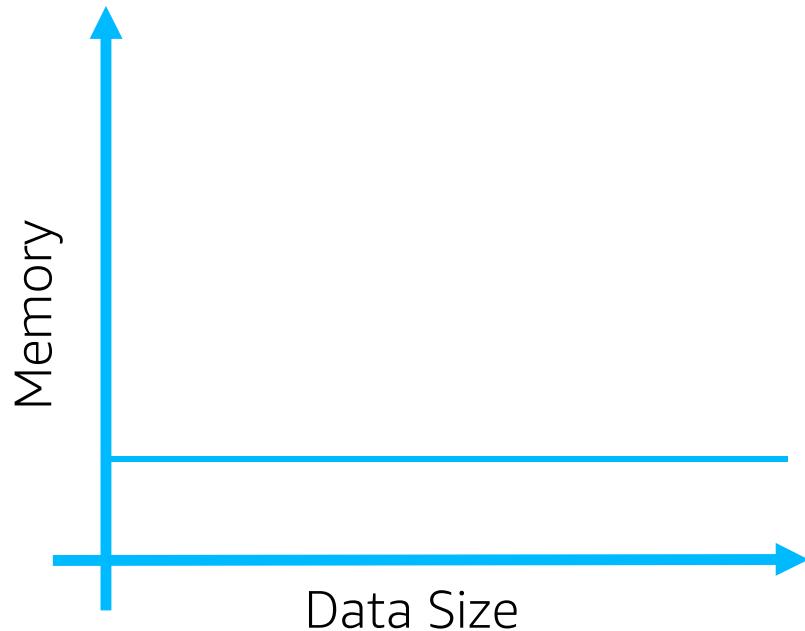
SageMaker Built-in Algorithms

Time vs. Money



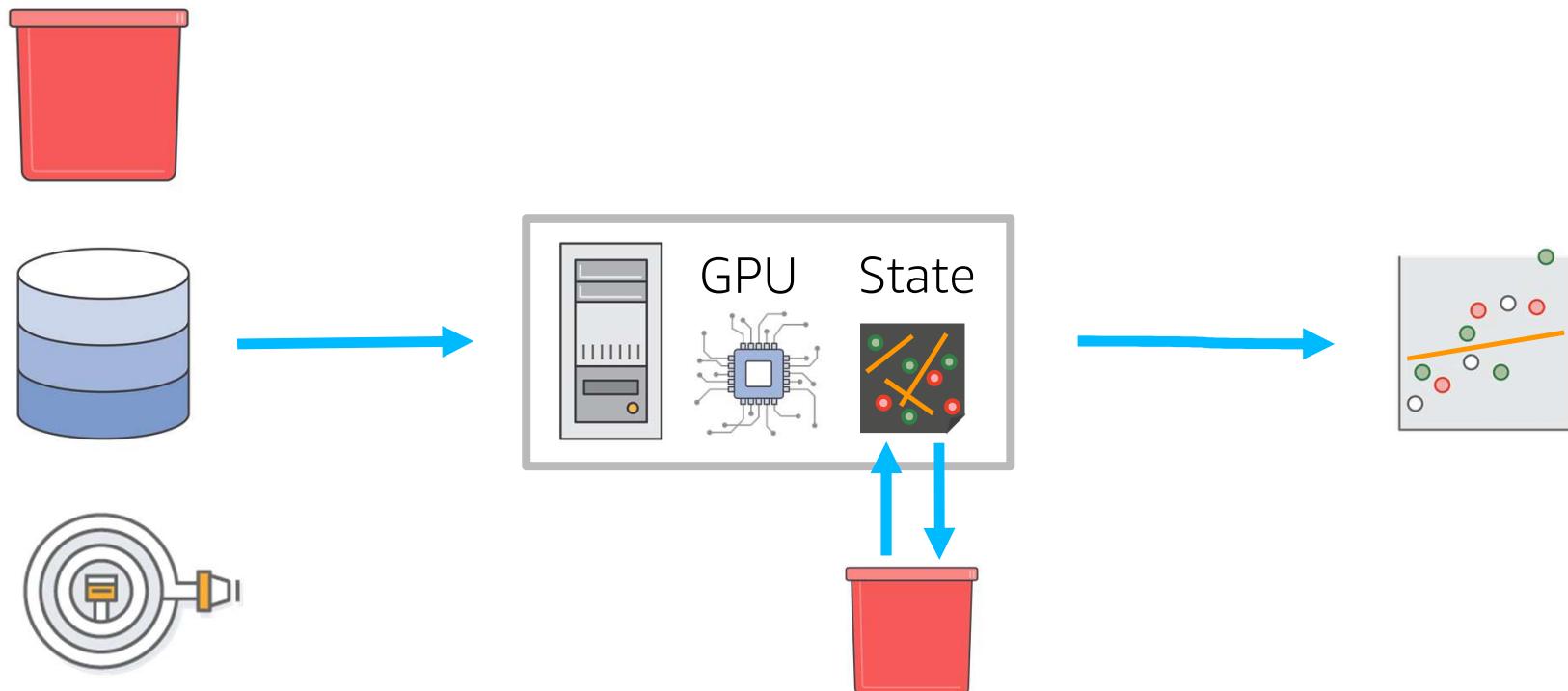
SageMaker Built-in Algorithms

Streaming



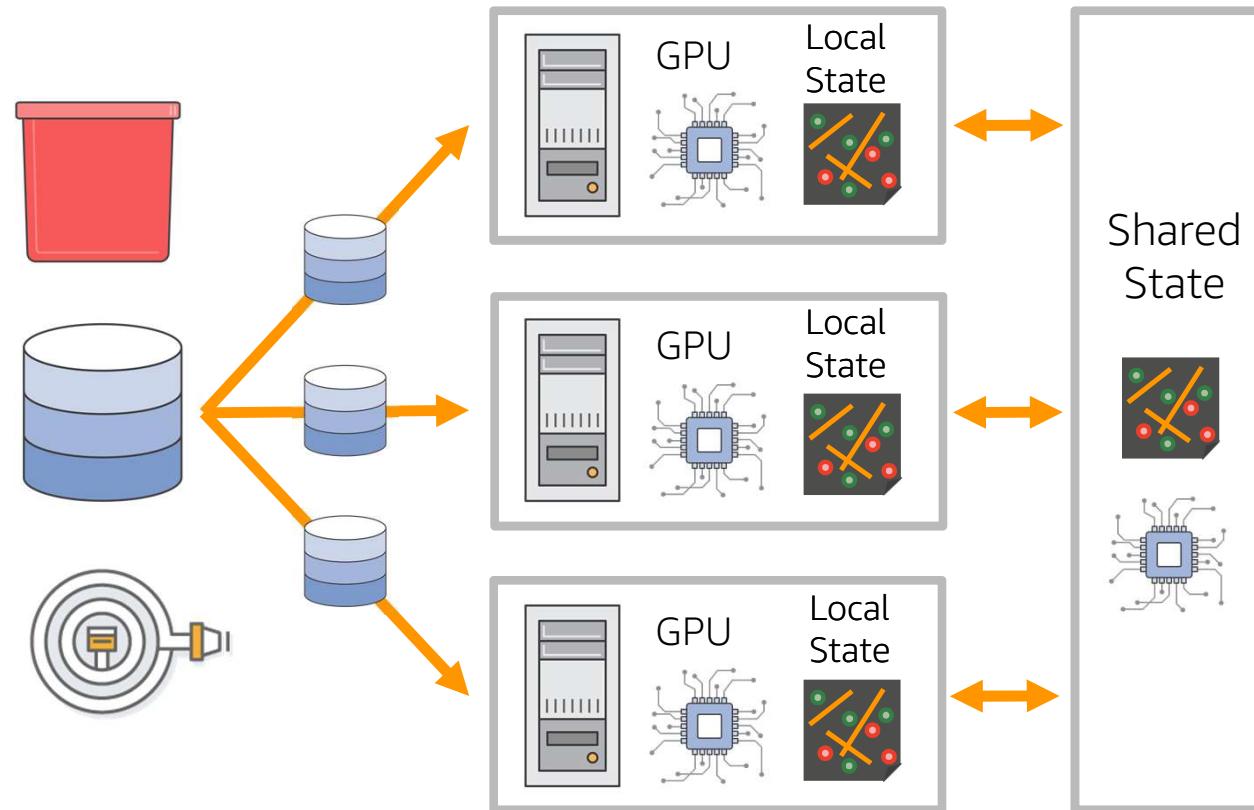
SageMaker Built-in Algorithms

Streaming



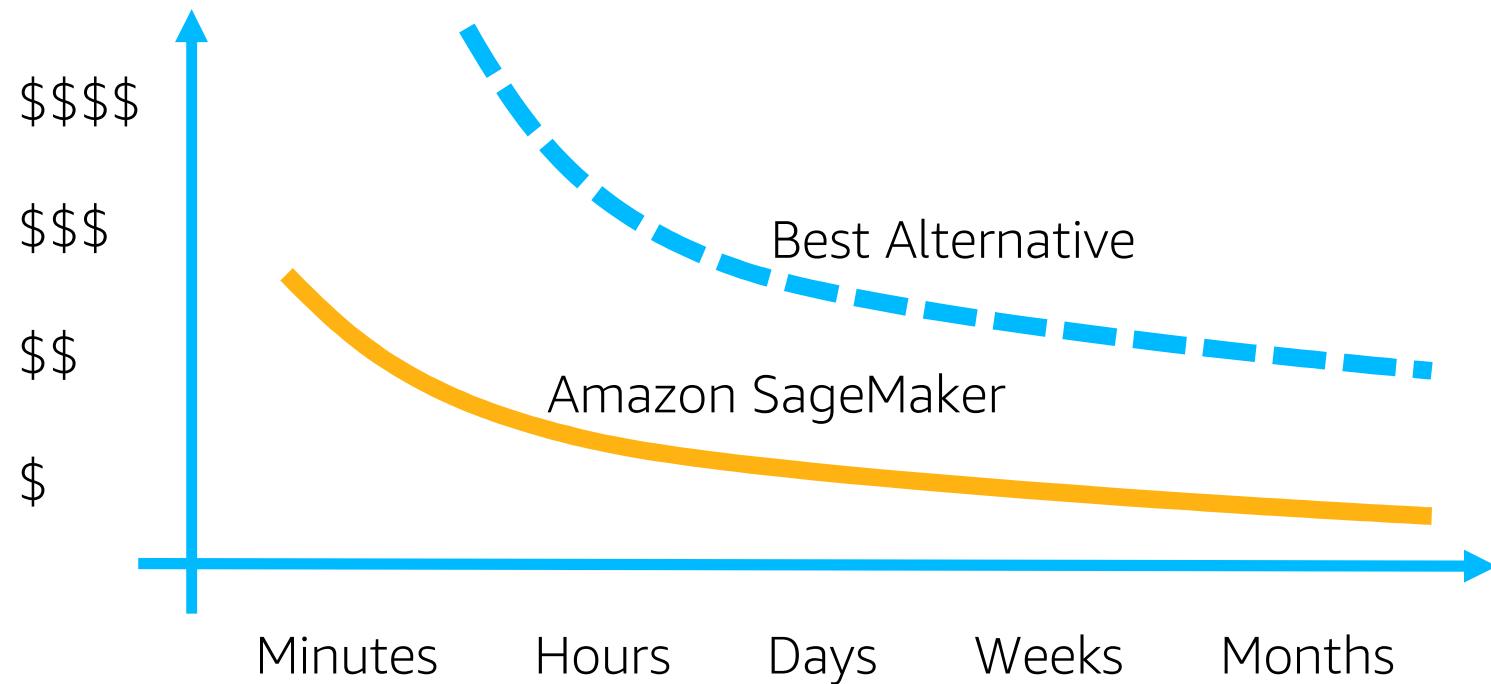
SageMaker Built-In Algorithms

Distributed Shared State



SageMaker Built-in Algorithms

Cost vs. Time





Infinitely Scalable ML Algorithms

Linear Learner



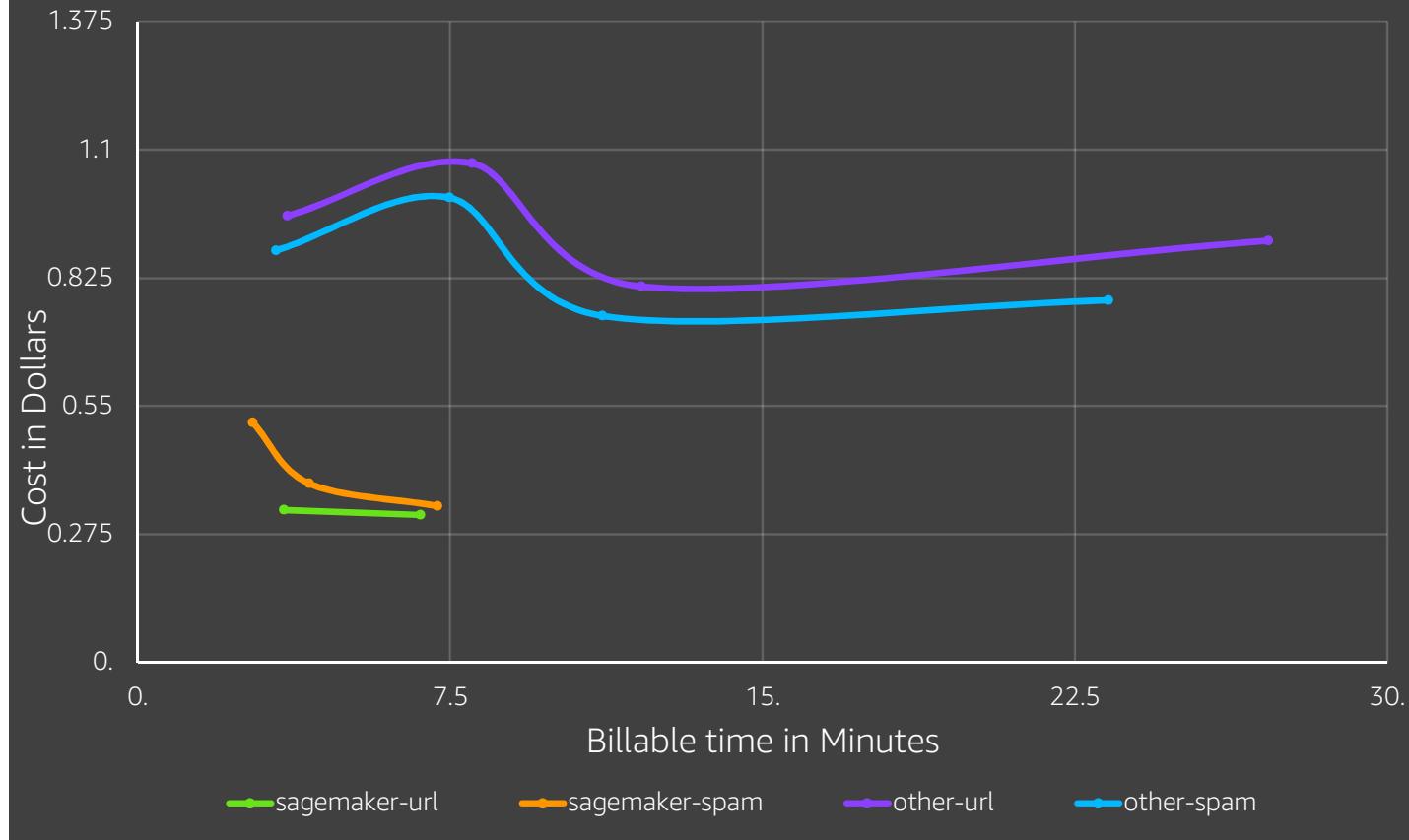
Regression (mean squared error)

SageMaker	Other
1.02	1.06
1.09	1.02
0.332	0.183
0.086	0.129
83.3	84.5

Classification (F1 Score)

SageMaker	Other
0.980	0.981
0.870	0.930
0.997	0.997
0.978	0.964
0.914	0.859
0.470	0.472
0.903	0.908
0.508	0.508

30 GB datasets for web-spam and web-url classification



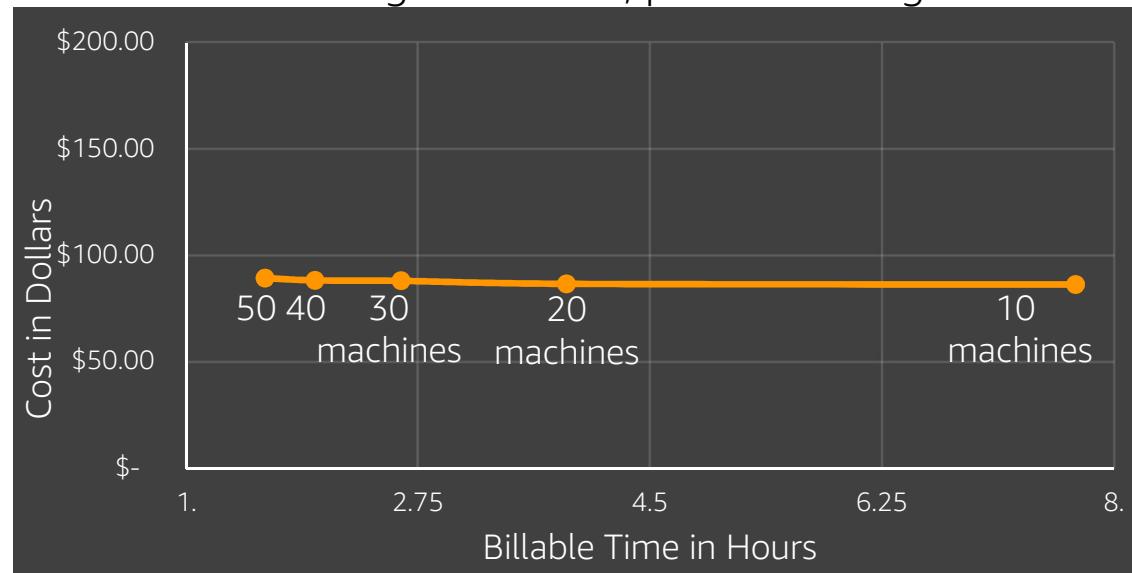
Factorization Machines



$$\tilde{y} = w_0 + \langle w_1, x \rangle + \sum_{i,j>i} x_i x_j \cdot \langle v_i, v_j \rangle$$

	Log_loss	F1 Score	Seconds
SageMaker	0.494	0.277	820
Other (10 Iter)	0.516	0.190	650
Other (20 Iter)	0.507	0.254	1300
Other (50 Iter)	0.481	0.313	3250

Click Prediction 1 TB advertising dataset,
m4.4xlarge machines, perfect scaling.



K-Means Clustering



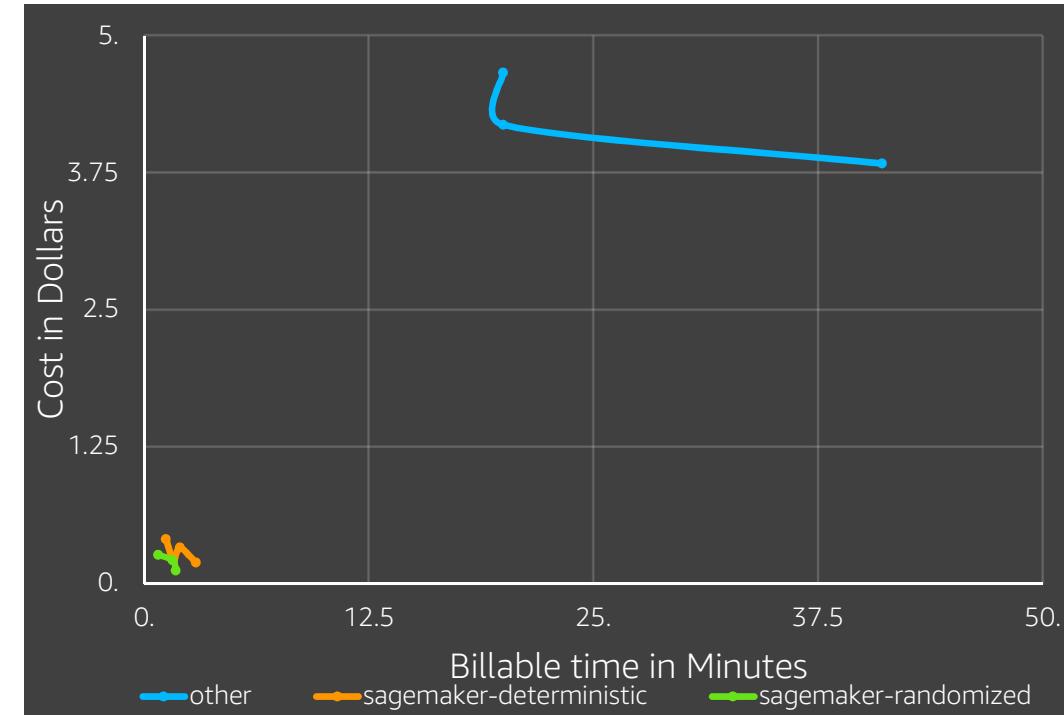
	k	SageMaker	Other
Text 1.2GB	10	1.18E3	1.18E3
	100	1.00E3	9.77E2
	500	9.18.E2	9.03E2
Images 9GB	10	3.29E2	3.28E2
	100	2.72E2	2.71E2
	500	2.17E2	Failed
Videos 27GB	10	2.19E2	2.18E2
	100	2.03E2	2.02E2
	500	1.86E2	1.85E2
Advertising 127GB	10	1.72E7	Failed
	100	1.30E7	Failed
	500	1.03E7	Failed
Synthetic 1100GB	10	3.81E7	Failed
	100	3.51E7	Failed
	500	2.81E7	Failed



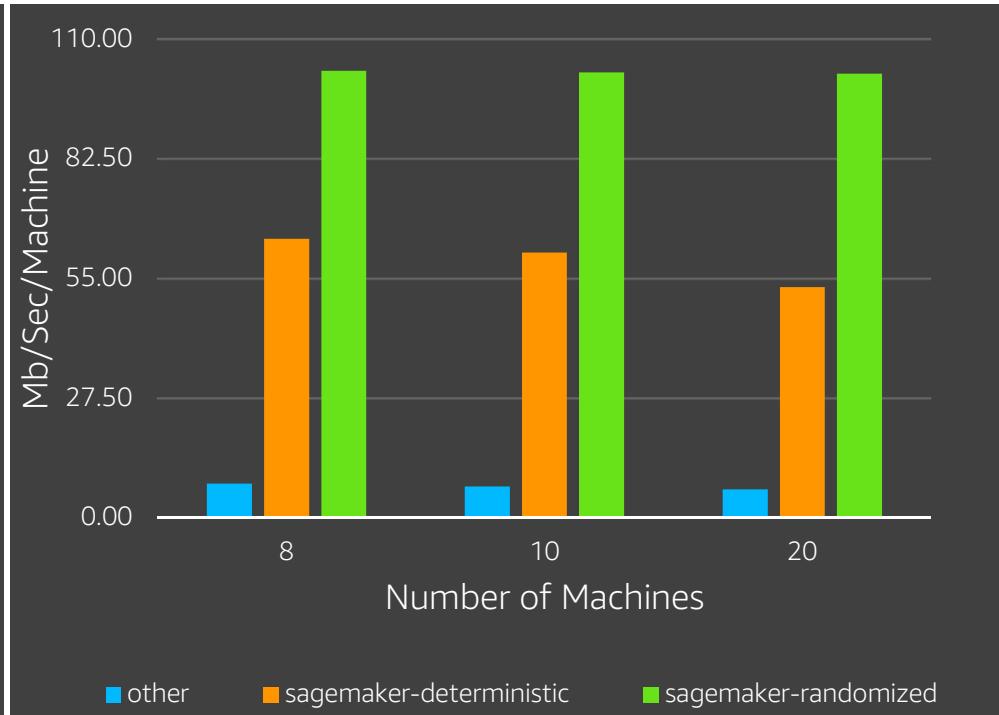
Principal Component Analysis (PCA)



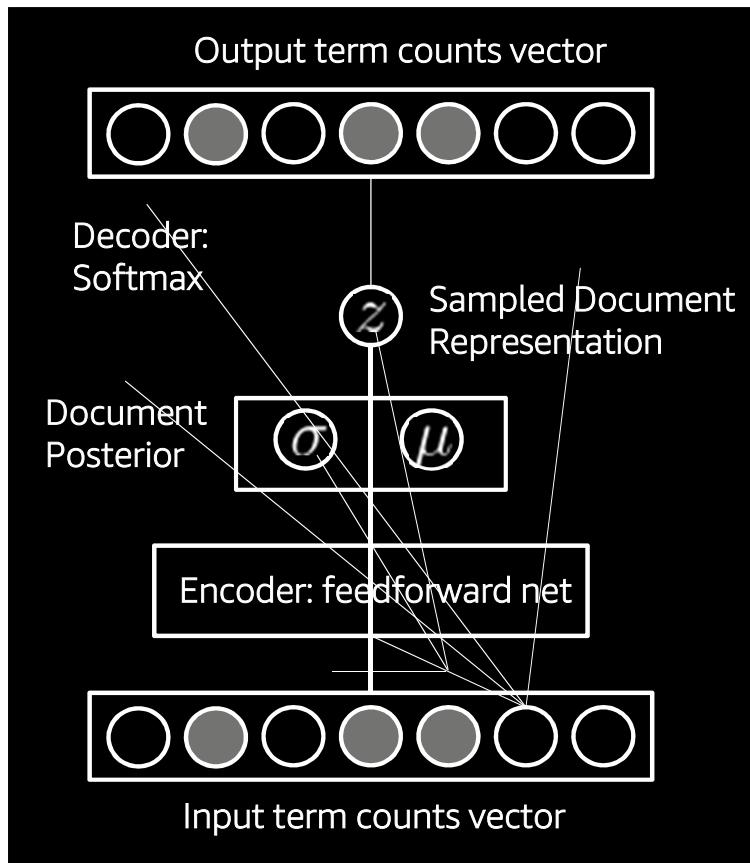
Cost vs. Time



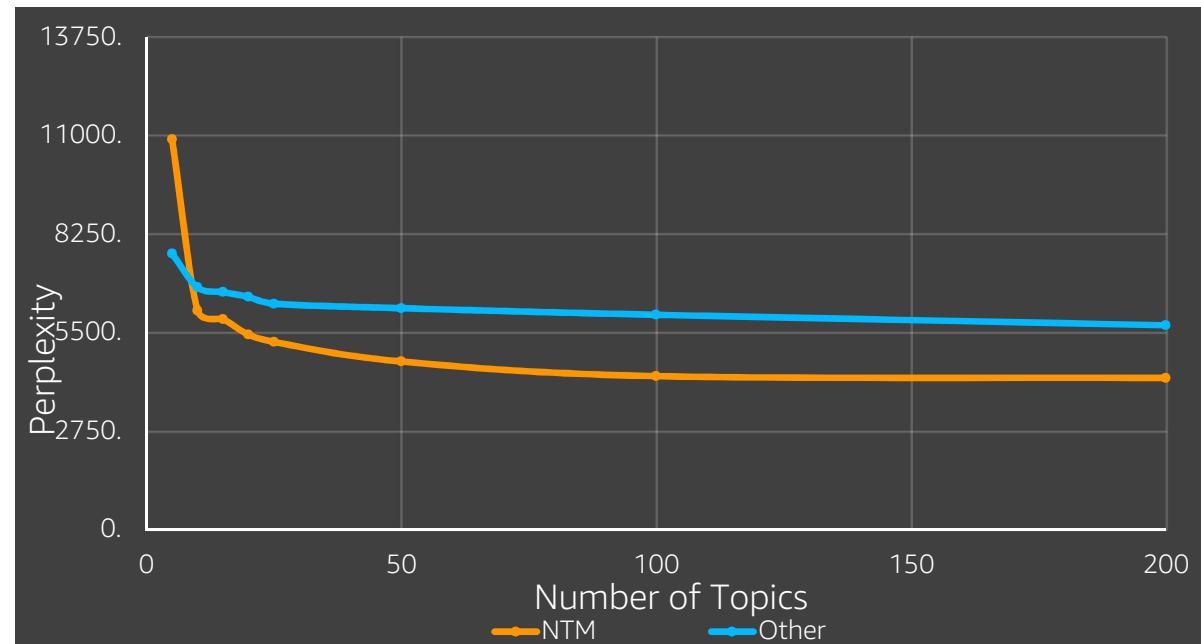
Throughput and Scalability



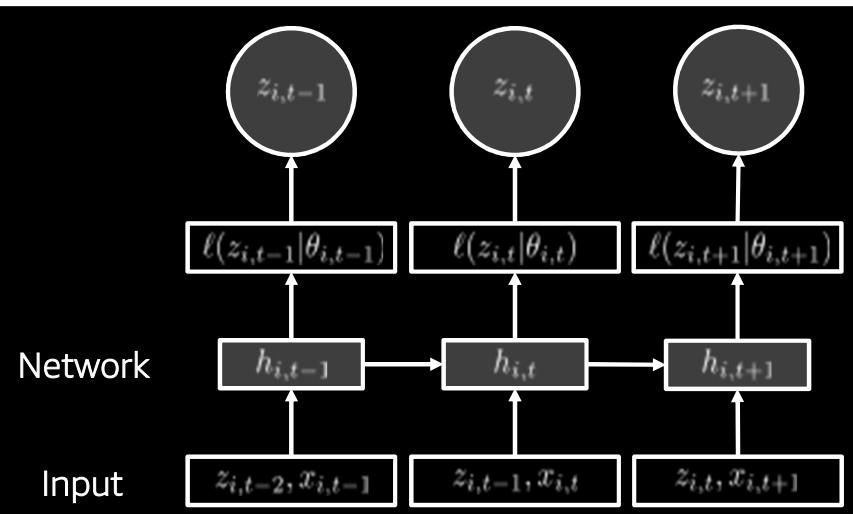
Neural Topic Modeling



Perplexity vs. Number of Topic
(~200K documents, ~100K vocabulary)



DeepAR



	Mean absolute percentage error		P90 Loss	
	DeepAR	R	DeepAR	R
Traffic Hourly occupancy rate of 963 bay area freeways	0.14	0.27	0.13	0.24
Electricity Electricity use of 370 homes over time	0.07	0.11	0.08	0.09
Page views Page view hits of websites	10k 180k	0.32 0.32	0.32 0.34	0.44 0.29
				NA

One hour on p2.xlarge, \$1





More Great ML Algorithms

Spectral LDA



The New York Times |

U.S.

High-Tech Industry, Long Shy of Politics, Is Now Belle of Ball

By LIZETTE ALVAREZ DEC. 26, 1999

Correction Appended

At a time when Congress is bitterly divided and unable to reach consensus on issues like gun control and health care, Democrats and Republicans are happily reaching across party lines to pass legislation backed by high-tech companies.

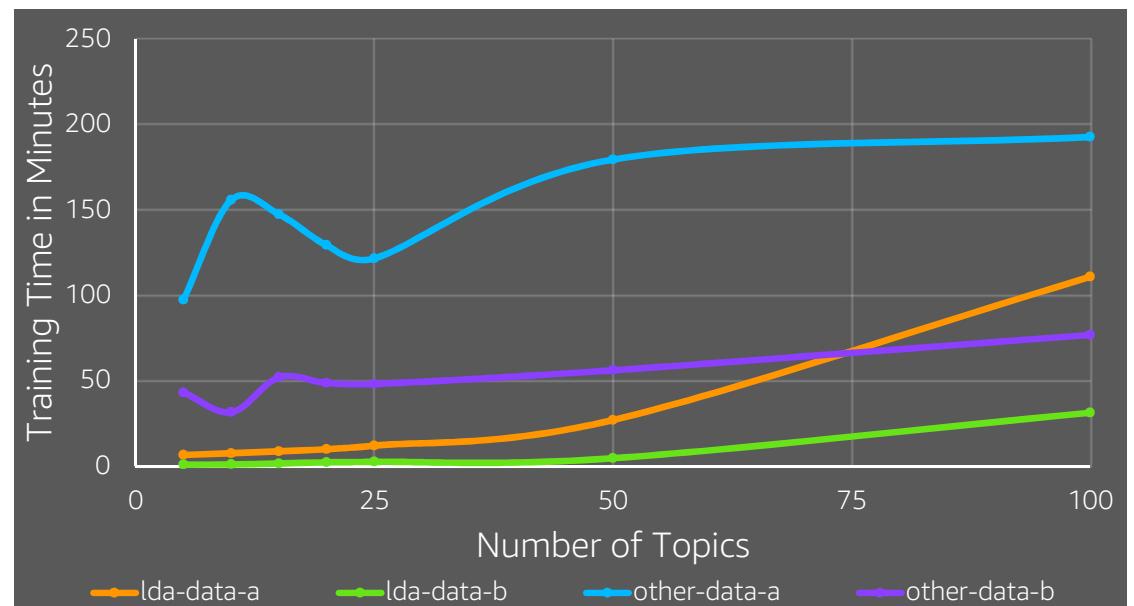
The high-tech industry, at the same moment, is lavishing new attention on Washington and changing its once-aloo posture toward the federal government.

Republicans and Democrats are both eager to win the loyalties of high-tech companies and executives, knowing that they represent untold jobs, wealth and ultimately votes and campaign contributions.

For its part, the industry has realized that the federal government can do its members as much harm as good. Microsoft, and its battle with the Justice Department, along with a spate of other threatened legal problems, drilled this point home.

"Microsoft was a poster child for our industry," said Connie Correll, director of communications for the Information Technology Industry Council, a trade organization that represents America Online, Dell and I.B.M., among others.

Training Time vs. Number of Topics



Boosted Decision Trees

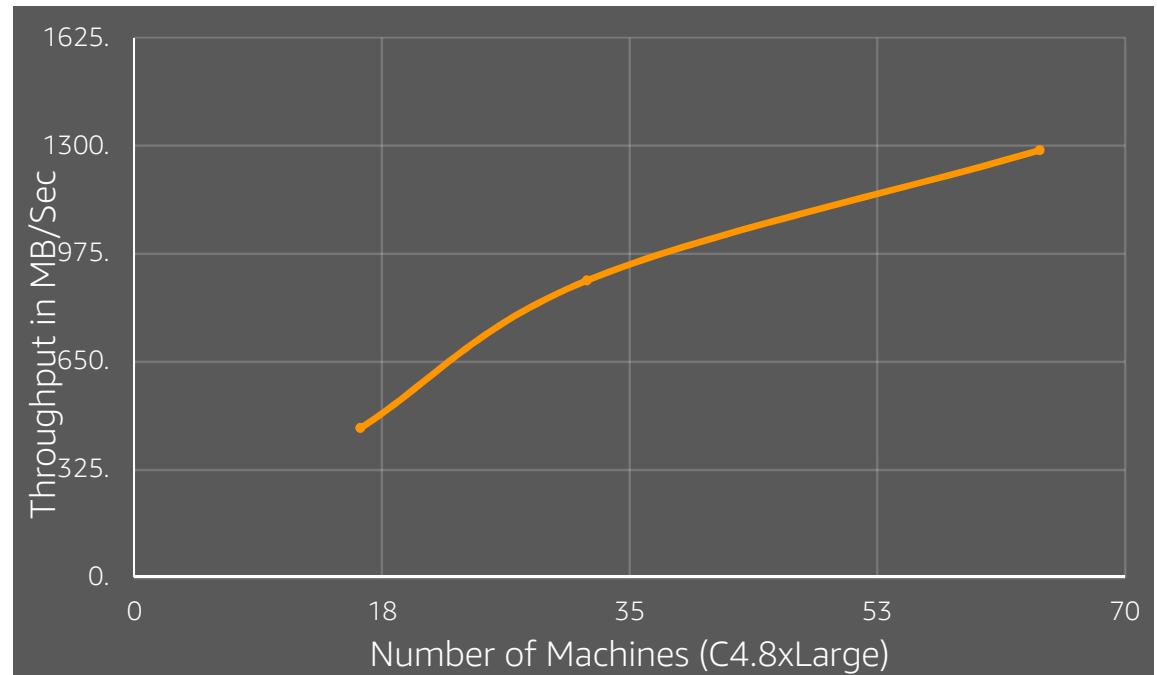


XGBoost is one of the most commonly used implementations of boosted decision trees in the world.

It is now available in Amazon SageMaker!

XGBoost

Throughput vs. Number of Machines



Sequence to Sequence



Based on Sockeye and Apache incubated MxNet, Multi-GPU, and can be used for Neural Machine Translation.

Supports both RNN/CNN as encoder/decoder

English-German Translation

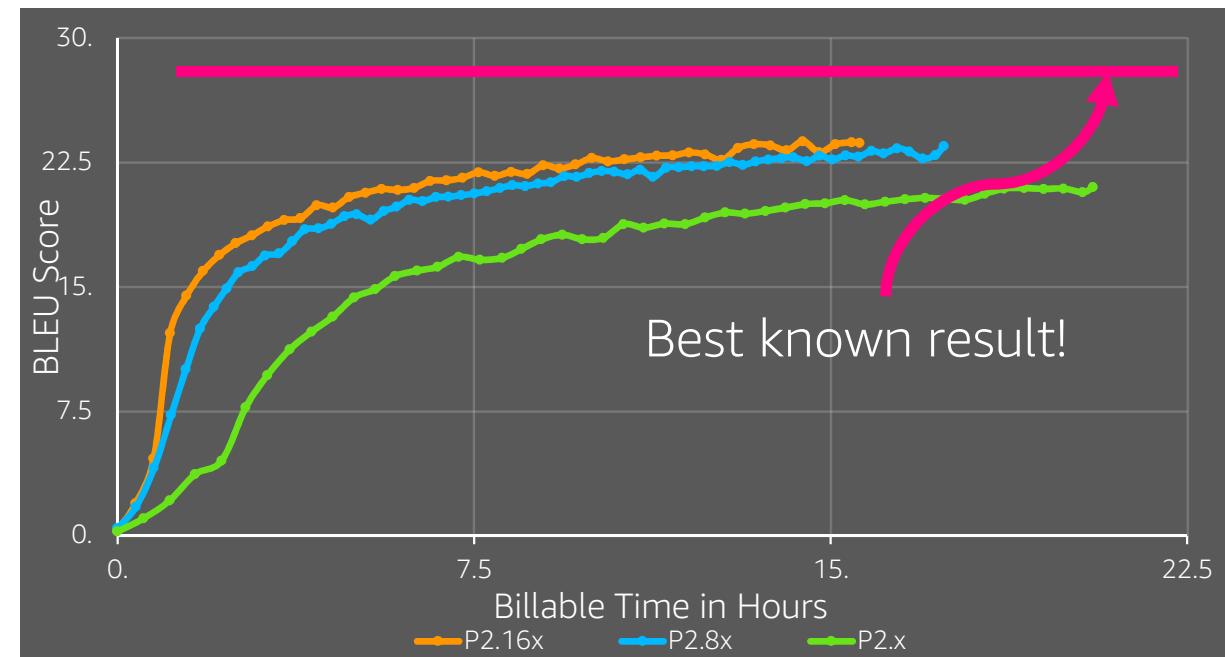


Image Classification

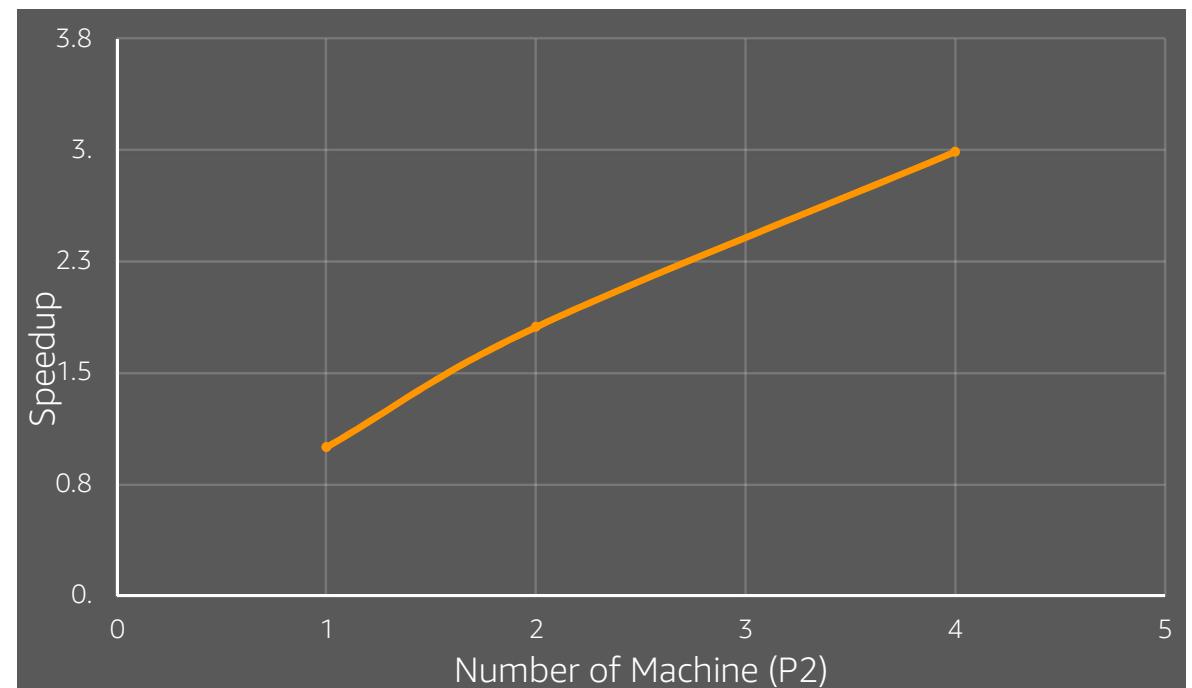


Implementation in MxNet of ResNet.

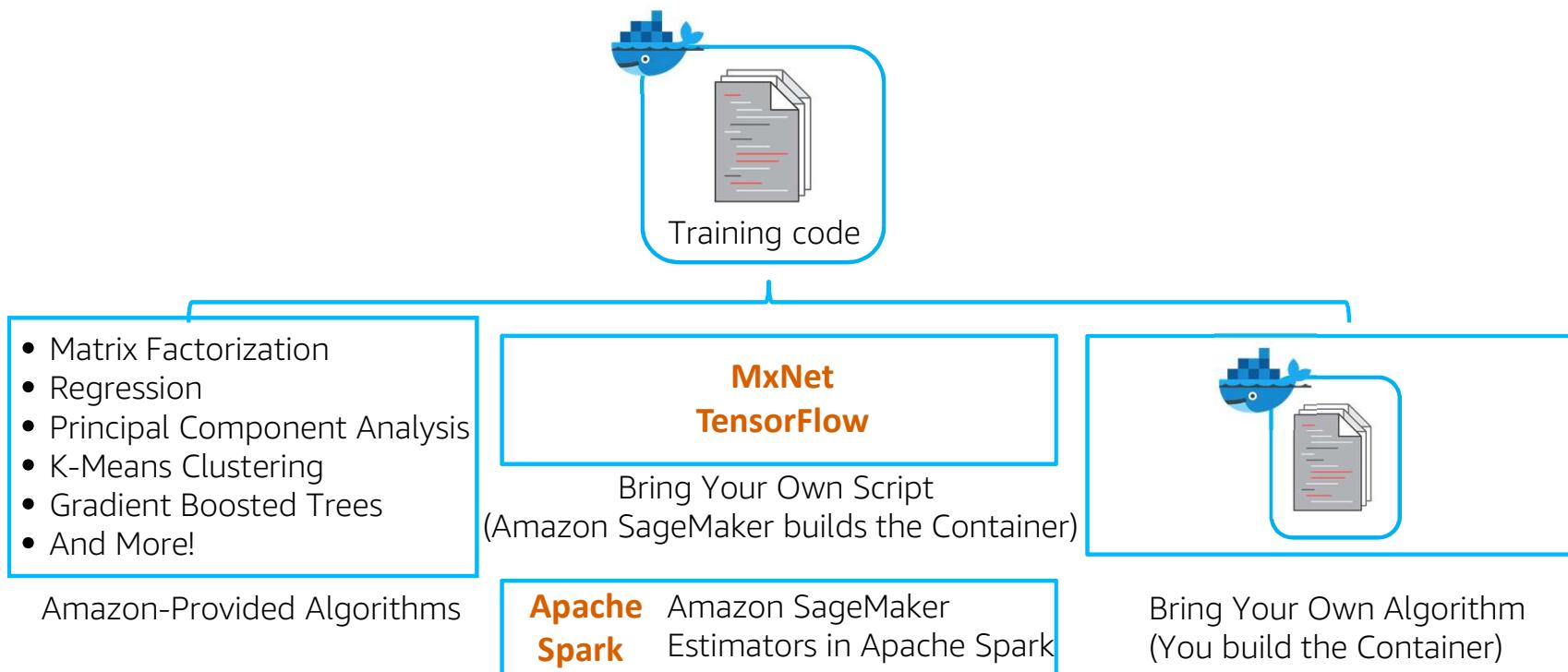
Other networks such as DenseNet and Inception will be added in the future.

Transfer learning: begin with a model already trained on ImageNet!

Speedup with Horizontal Scaling

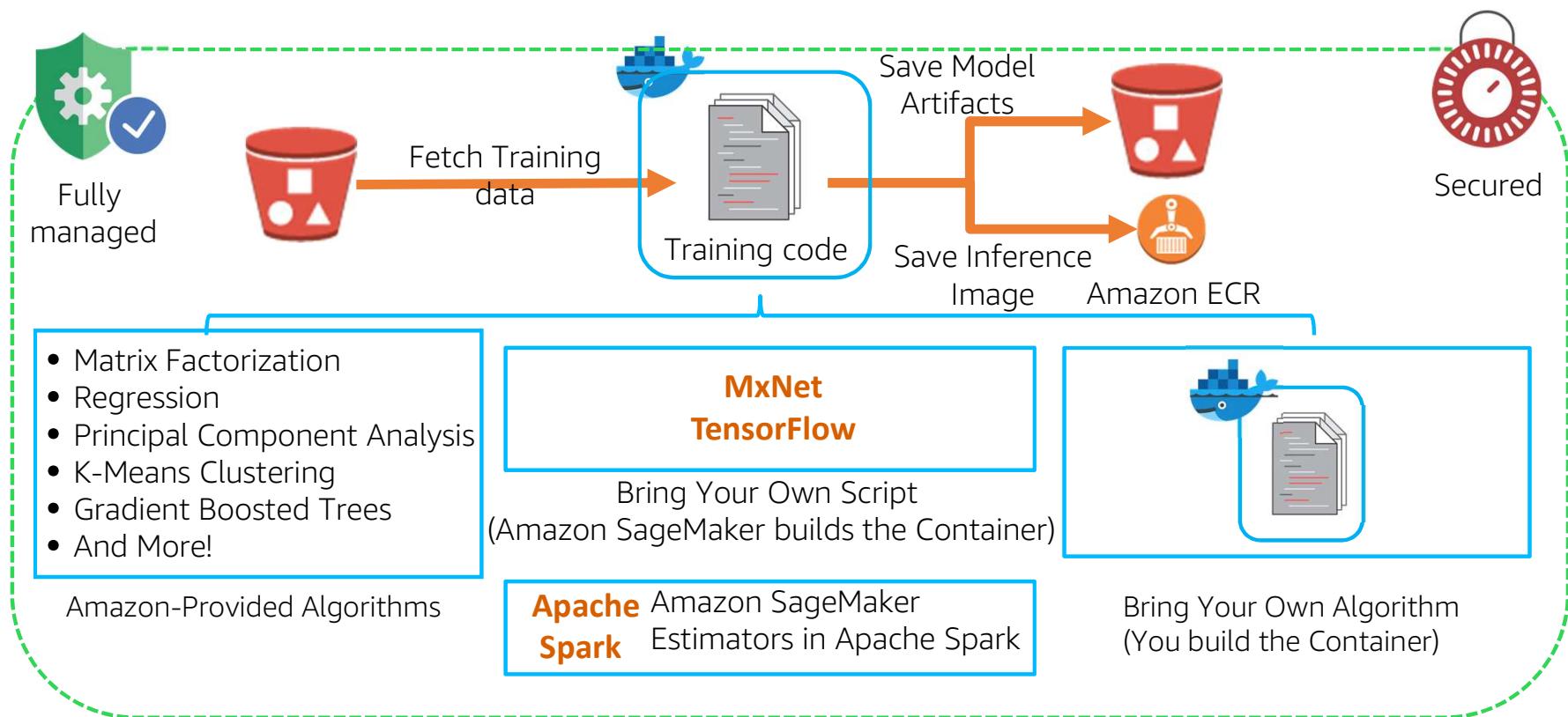


Amazon SageMaker Built-In Algorithms 10x Better



Amazon SageMaker Built-In Algorithms

Managed Distributed Training with Flexibility





Demo 6: Using Amazon SageMaker Built-in Algorithms

Amazon SageMaker Hosting Service

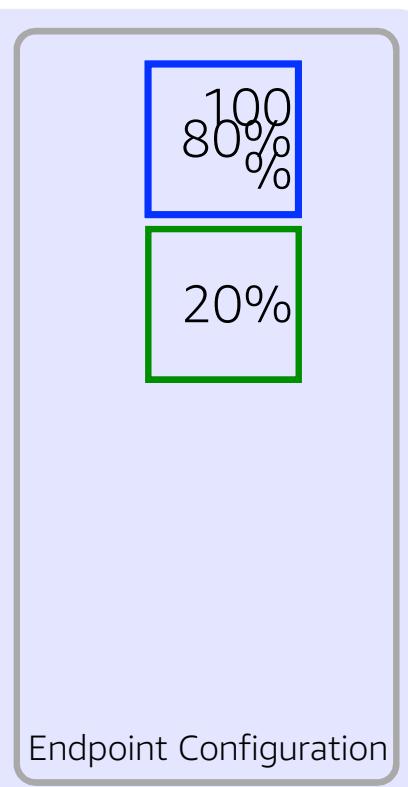
Easy Model Deployment to Amazon SageMaker



- InstanceType: c3.4xlarge
- InitialInstanceCount: 3
- ModelName: prod
- VariantName: primary
- InitialVariantWeight: 100



Inference
EndPoint



SageMaker Hosting Service

Easy Model Deployment to Amazon SageMaker



- Auto-Scaling Inference APIs
- A/B Testing (more to come)
- Low Latency & High Throughput
- Bring Your Own Model
- Python SDK

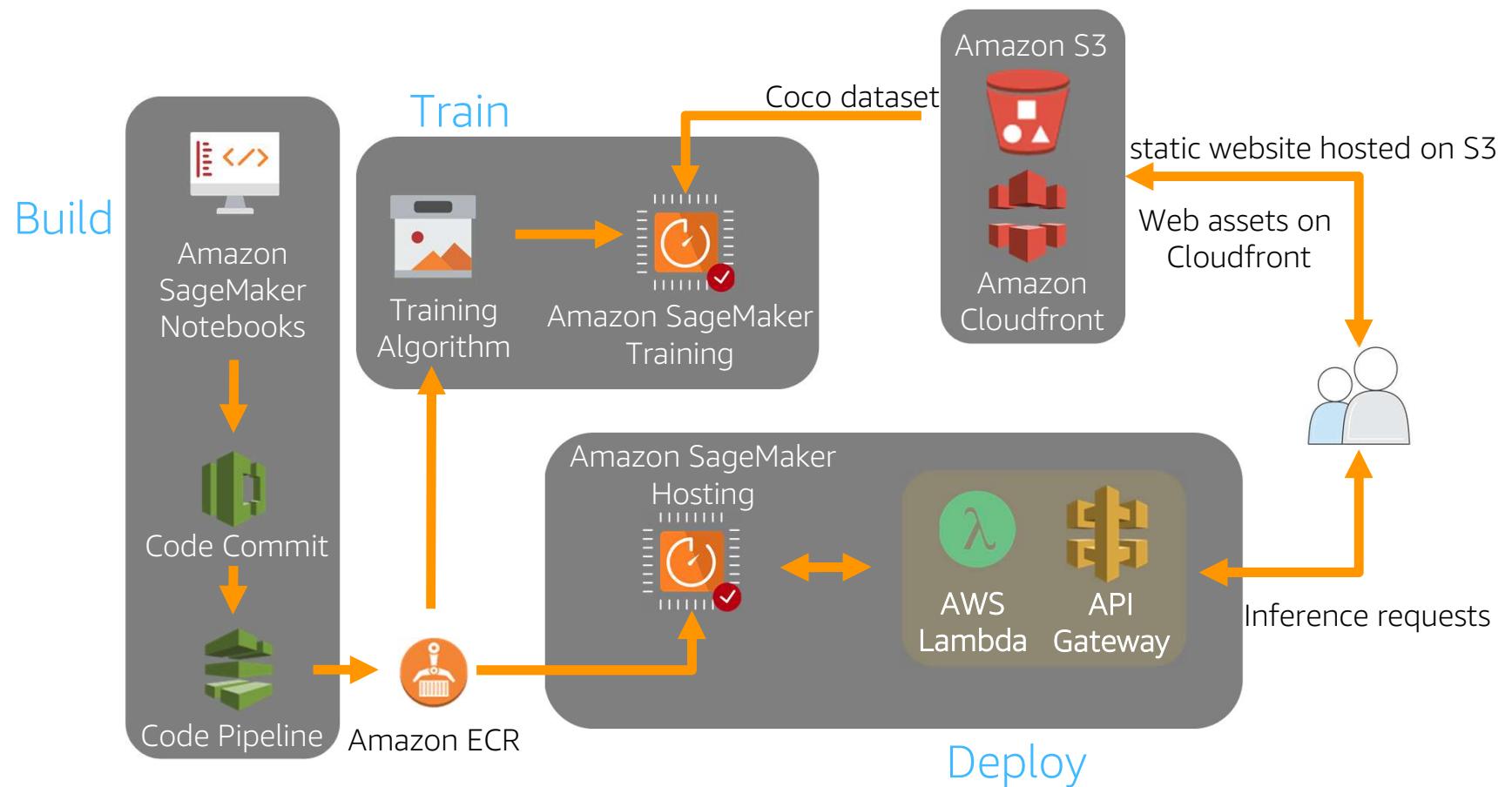


Demo 7: Analyzing Breast Cancer Datasets



Demo 8: Using Containers with Amazon SageMaker

Amazon SageMaker Reference Architecture



Amazon SageMaker Technology Competency Partners



Data Services	Platform Solutions		SaaS and API Solutions	
Alteryx	Bonsai	DataRobot	Anodot	SigOpt
CrowdFlower	C3 IoT	DOMINO DATA LAB	Luminoso	Veritone
Paxata	Databricks	H2O.ai	Narrative Science	x.ai
TRIFACTA	Data Iku			

Call To Action



- Getting started with Amazon SageMaker:
<https://aws.amazon.com/sagemaker/>
- Use the Amazon SageMaker SDK:
 - For Python: <https://github.com/aws/sagemaker-python-sdk>
 - For Spark: <https://github.com/aws/sagemaker-spark>
- SageMaker Examples: <https://github.com/awslabs/amazon-sagemaker-examples>

Thank You

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