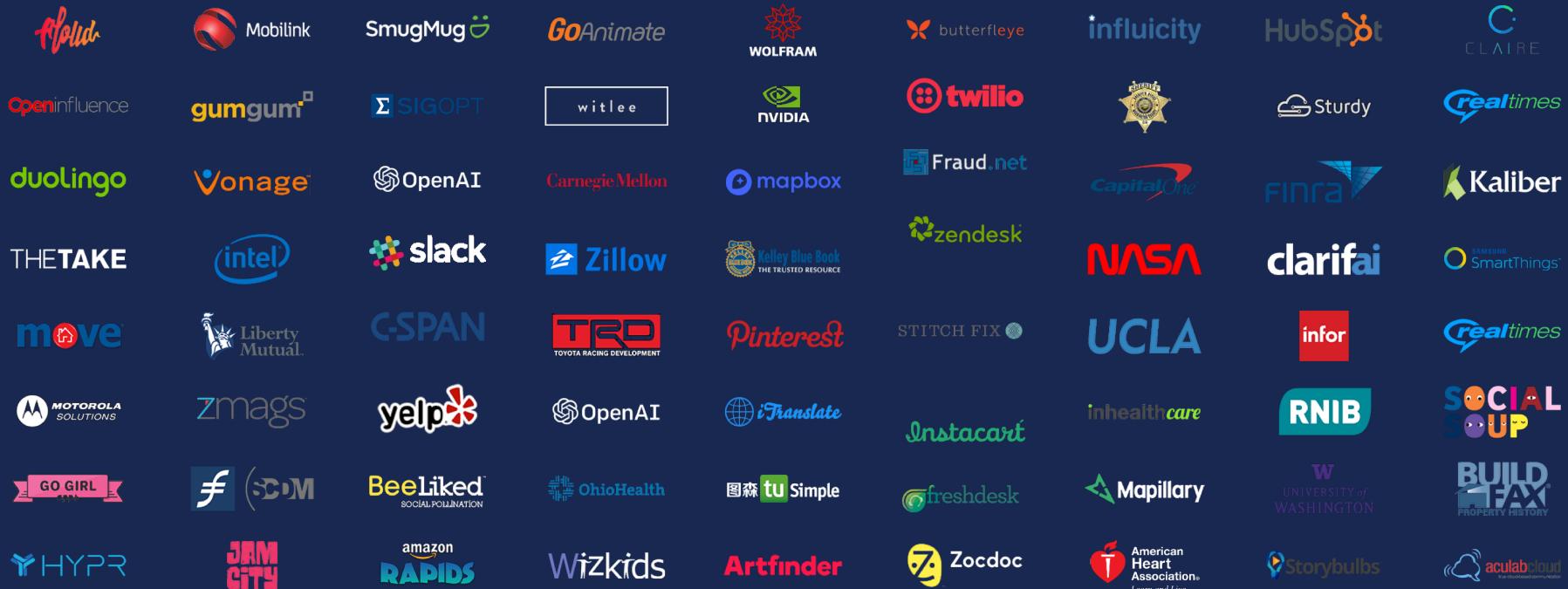


Recommender Workshop

Part 1: Introduction

Customers Running Machine Learning On AWS Today

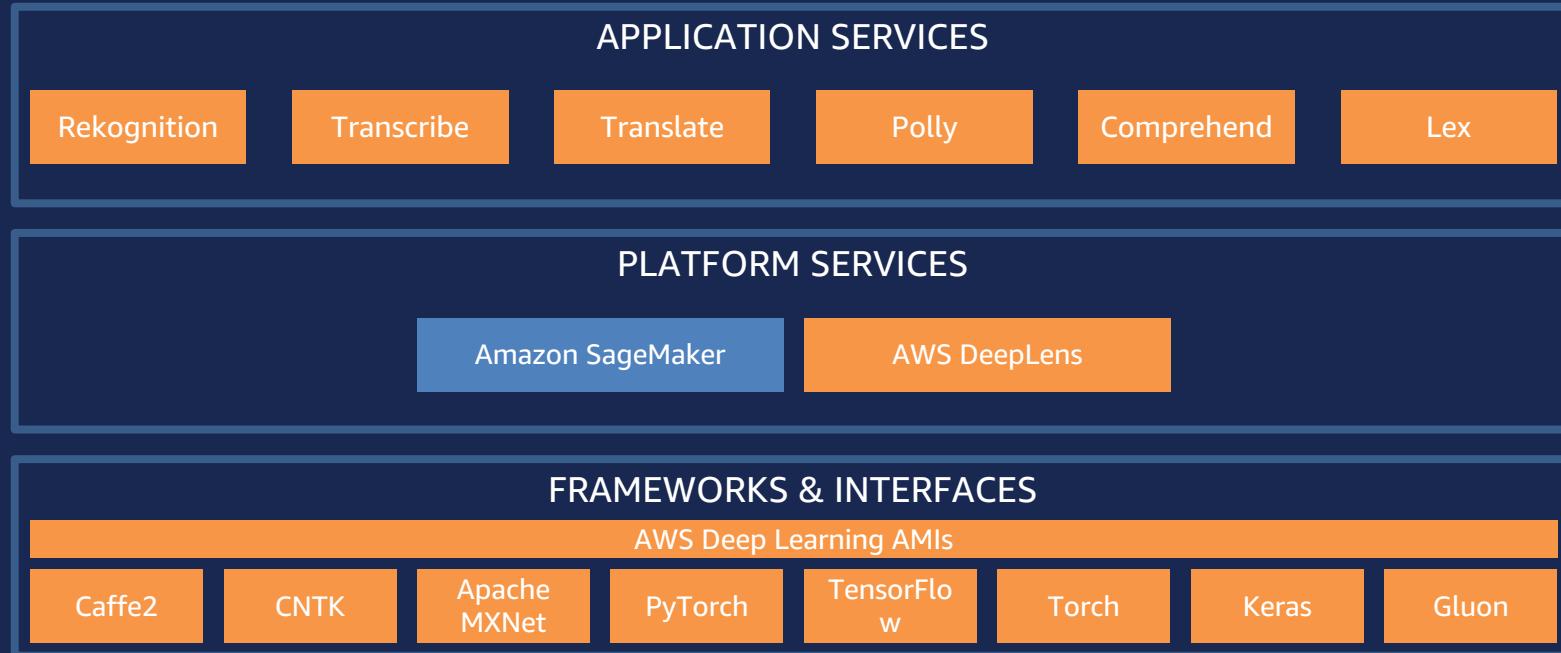


Machine Learning at AWS

Our mission:

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

The Amazon Machine Learning Stack

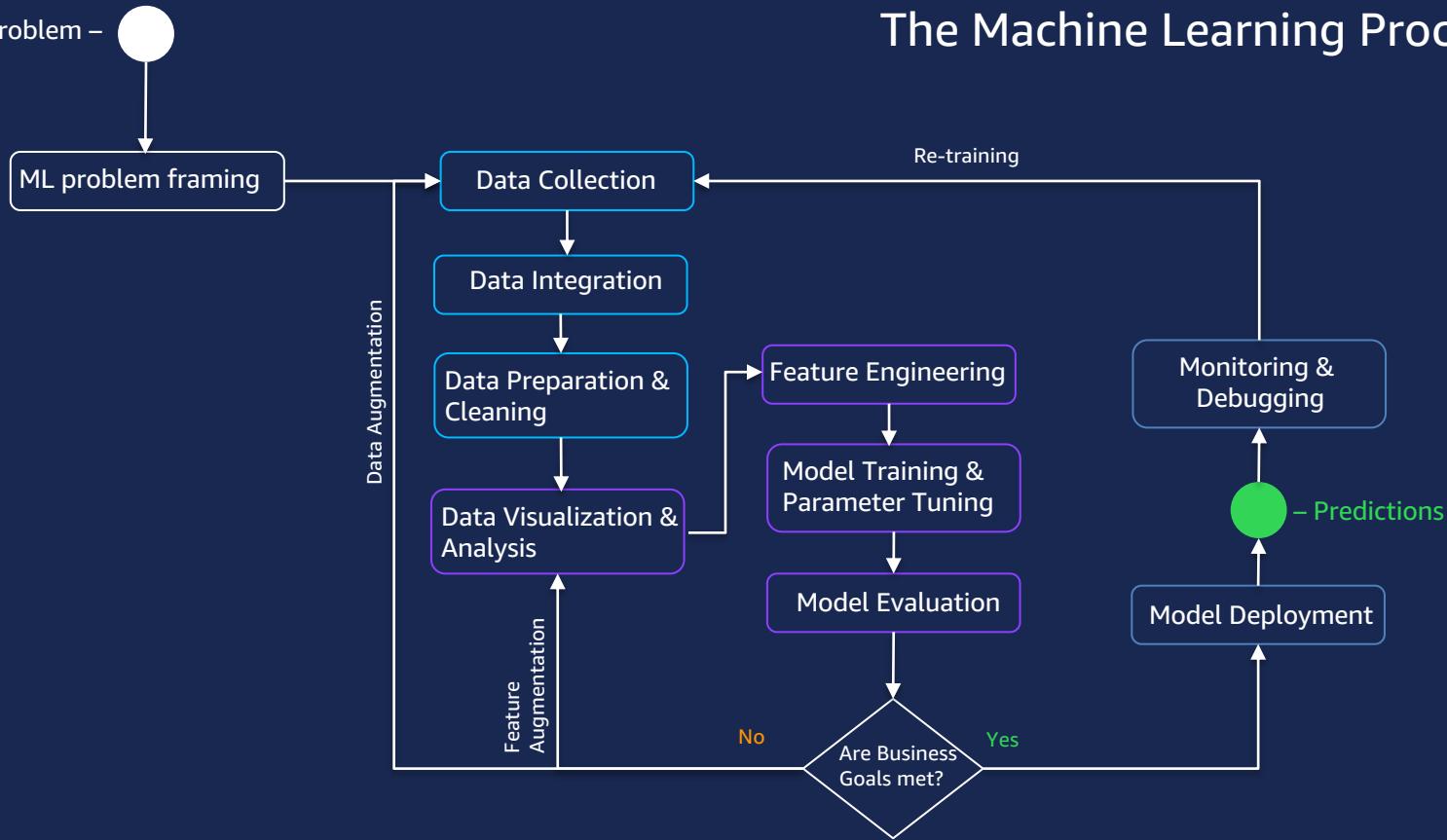


Recommender Workshop Agenda

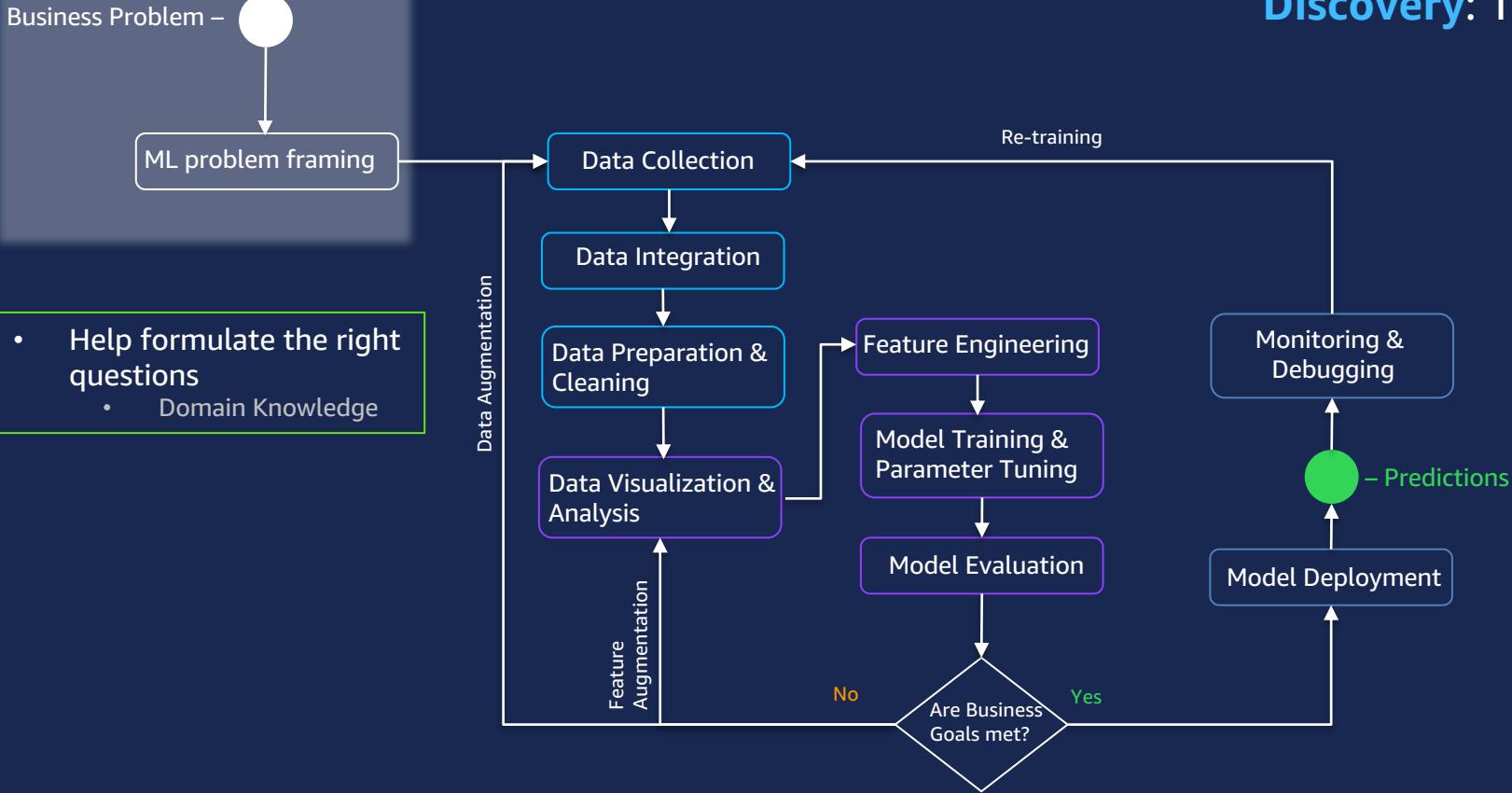
- Part 1: Introduction (You Are Here)
 - Overview of Machine Learning Process, Amazon SageMaker
 - Hands-on: **Data Exploration**
- Part 2: Collaborative Filtering
 - Core Concepts for Recommendations
 - Hands-on: **K-Means Clustering**
- Part 3: Matrix Factorization
 - Refining Recommendations
 - Hands-on: **Factorization Machine**
- Part 4: Hyperparameter Optimization

Let's Review the ML Process

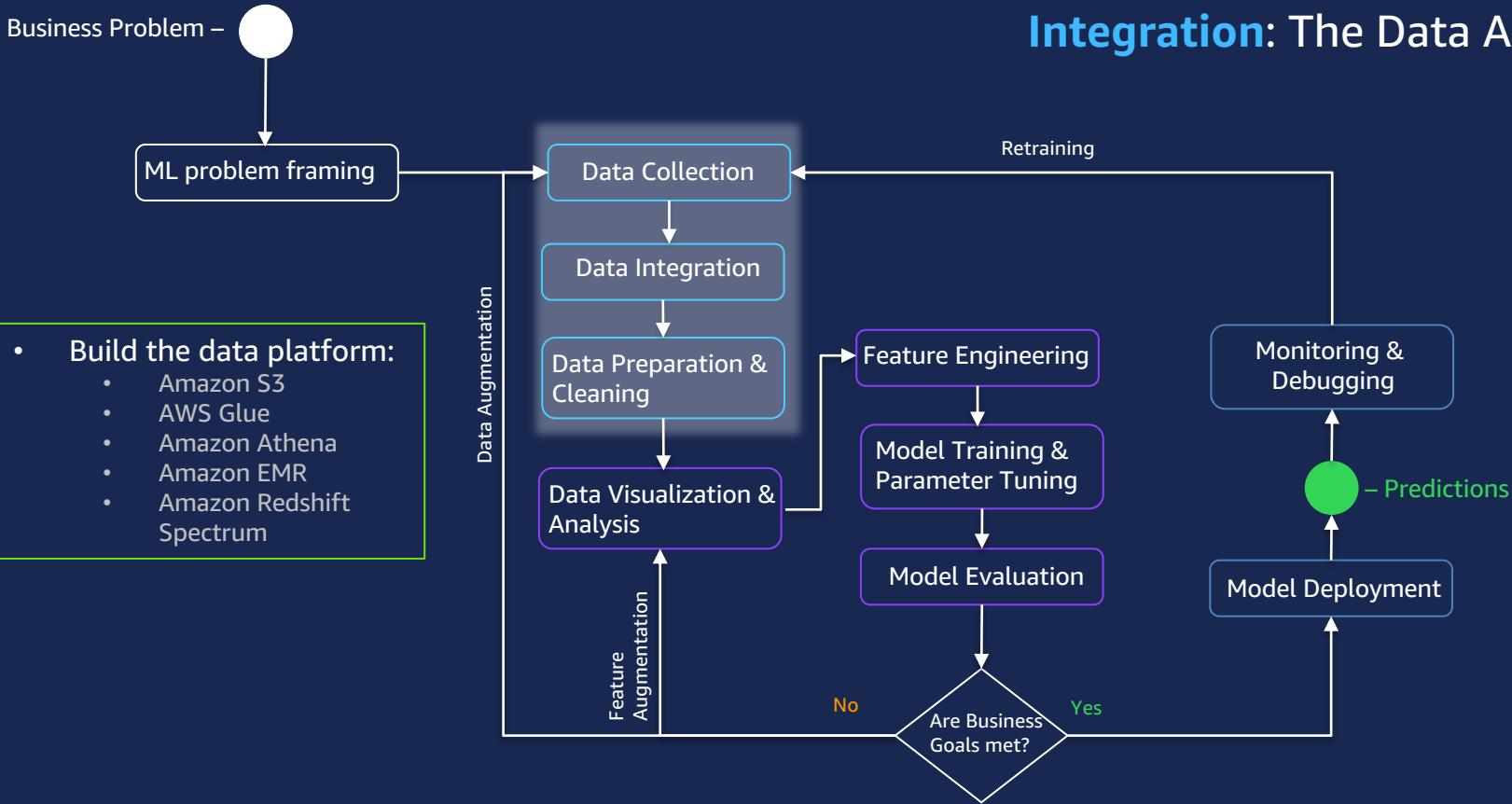
The Machine Learning Process



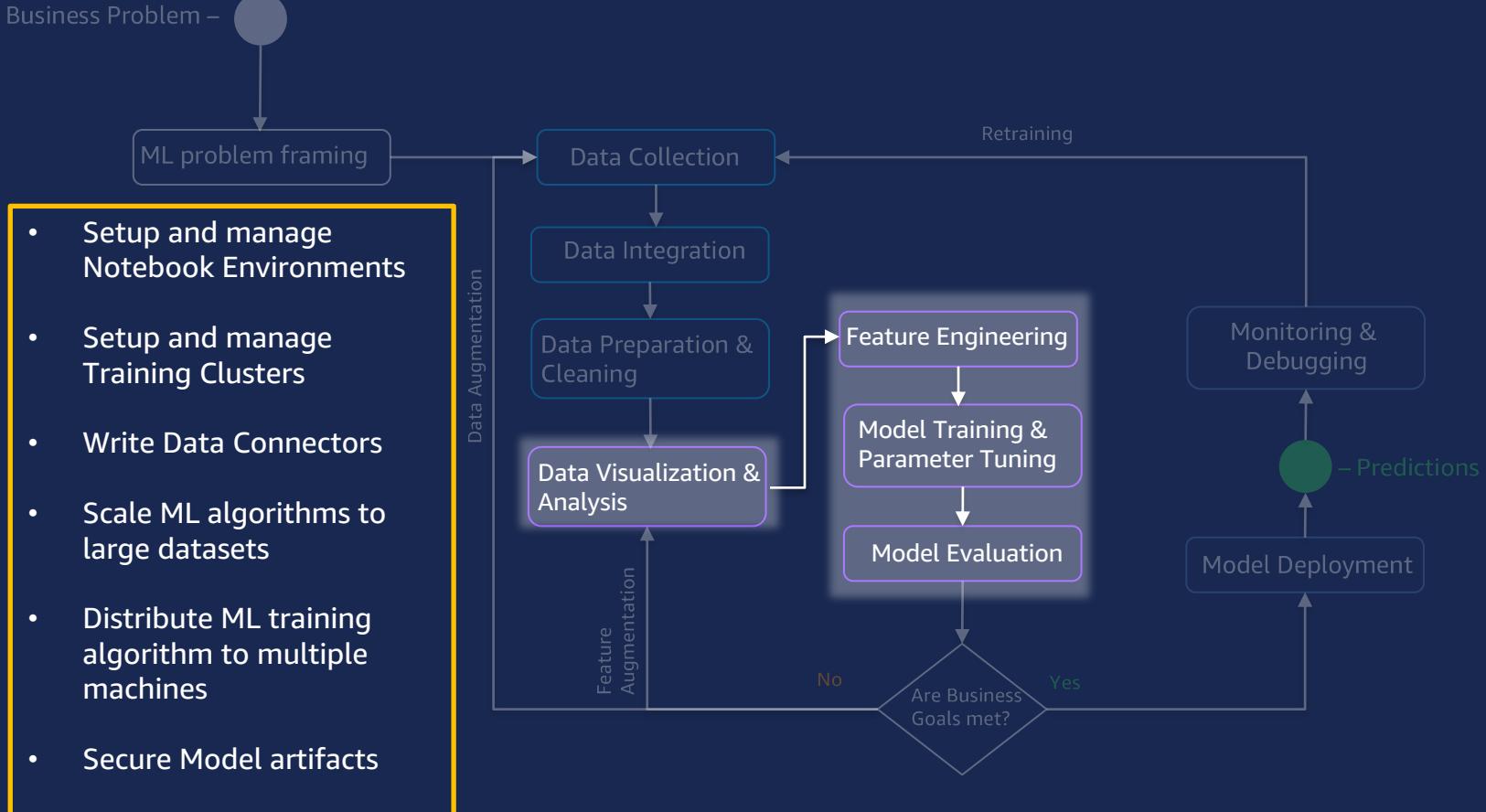
Discovery: The Analysts



Integration: The Data Architecture



Why We built Amazon SageMaker: The Model Training Undifferentiated Heavy Lifting

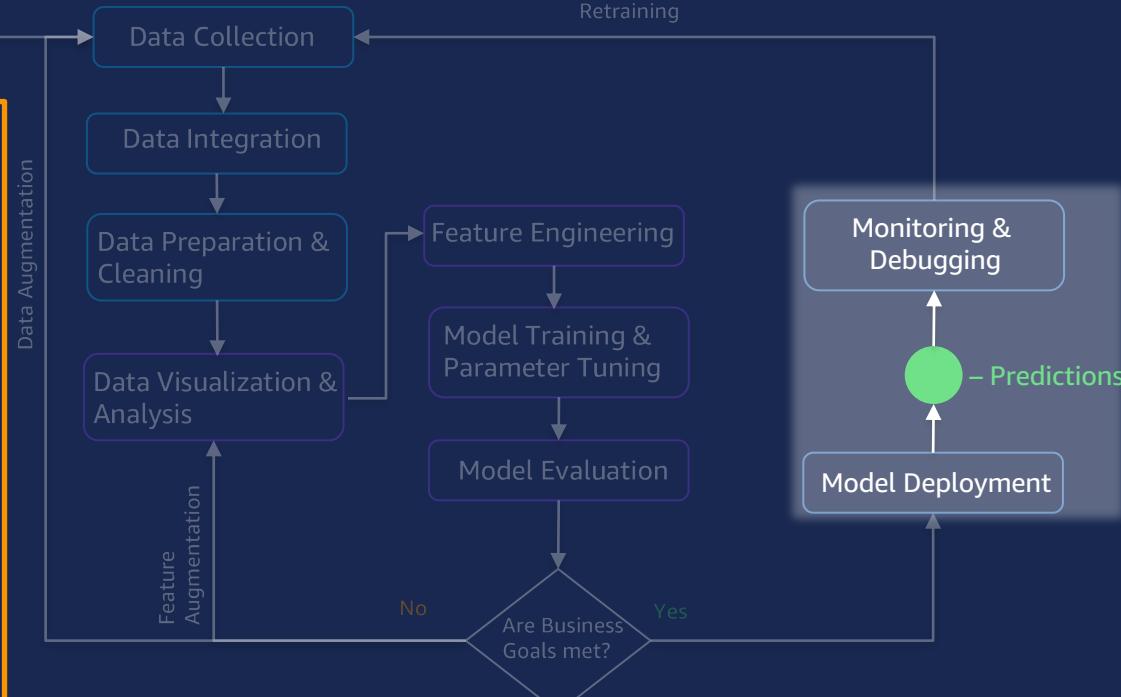


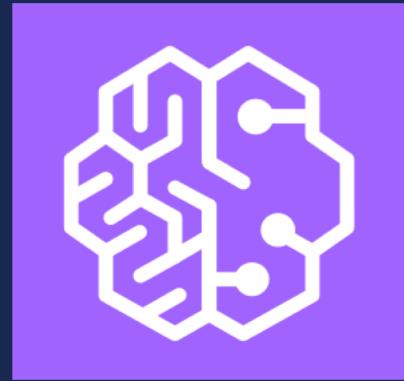
Why We built Amazon SageMaker: The Model Deployment Undifferentiated Heavy Lifting

Business Problem –



- Setup and manage Model 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)

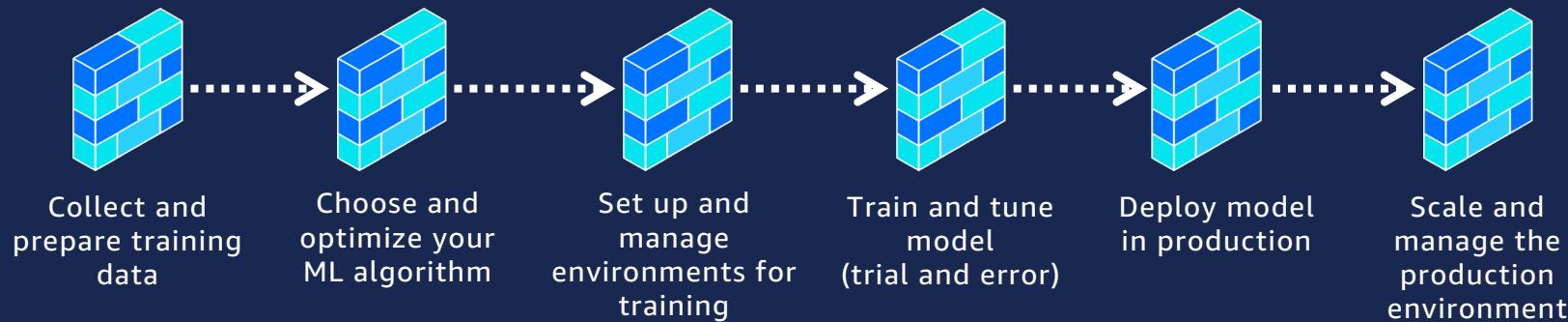




Amazon SageMaker

Amazon SageMaker

Easily build, train, and deploy machine learning models



Amazon SageMaker

Easily build, train, and deploy machine learning models



Pre-built
notebooks
for common
problems



Built-in, high
performance
algorithms



Set up and
manage
environments
for training



Train and
tune model
(trial and
error)



Deploy model
in production



Scale and
manage the
production
environment

BUILD

Training ML Models Using Amazon SageMaker

SageMaker Built-in Algorithms

- K-means Clustering
- PCA
- Neural Topic Modelling
- Factorisation Machines
- Linear Learner
- XGBoost
- Latent Dirichlet Allocation
- Image Classification
- Seq2Seq
- DeepAR Forecasting
- BlazingText (word2vec)
- Random Cut Forest

Training ML Models Using Amazon SageMaker

SageMaker Built-in Algorithms

K-means Clustering
PCA
Neural Topic Modelling
Factorisation Machines
Linear Learner – Regression
XGBoost
Latent Dirichlet Allocation
Image Classification
Seq2Seq
Linear Learner – Classification
DeepAR Forecasting

Bring Your Own Algorithms

ML Algorithms
R
MXNet
TensorFlow
Caffe
PyTorch
Keras
CNTK
...



Training ML Models Using Amazon SageMaker

SageMaker Built-in Algorithms

K-means Clustering
PCA
Neural Topic Modelling
Factorisation Machines
Linear Learner – Regression
XGBoost
Latent Dirichlet Allocation
Image Classification
Seq2Seq
Linear Learner – Classification
DeepAR Forecasting

Bring Your Own Algorithms

ML Algorithms
R
MXNet
TensorFlow
Caffe
PyTorch
Keras
CNTK
...

MXNet & TensorFlow SDK

TensorFlow SDK
MXNet (Gluon) SDK



Training ML Models Using Amazon SageMaker

SageMaker Built-in Algorithms

K-means Clustering
PCA
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Linear Learner – Regression
XGBoost
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DeepAR Forecasting

Bring Your Own Algorithms

ML Algorithms
R
MXNet
TensorFlow
Caffe
PyTorch
Keras
CNTK
...

MXNet & TensorFlow SDK

TensorFlow SDK
MXNet (Gluon) SDK

Apache Spark Estimator

Apache Spark Python library
Apache Spark Scala library



Amazon SageMaker

Easily build, train, and deploy machine learning models



Pre-built
notebooks
for common
problems



Built-in, high
performance
algorithms

BUILD



One-click
training



Hyperparameter
optimization



Deploy model
in production



Scale and
manage the
production
environment

TRAIN

Amazon SageMaker

Easily build, train, and deploy machine learning models



Pre-built
notebooks
for common
problems



Built-in, high
performance
algorithms



One-click
training



Hyperparameter
optimization



One-click
deployment



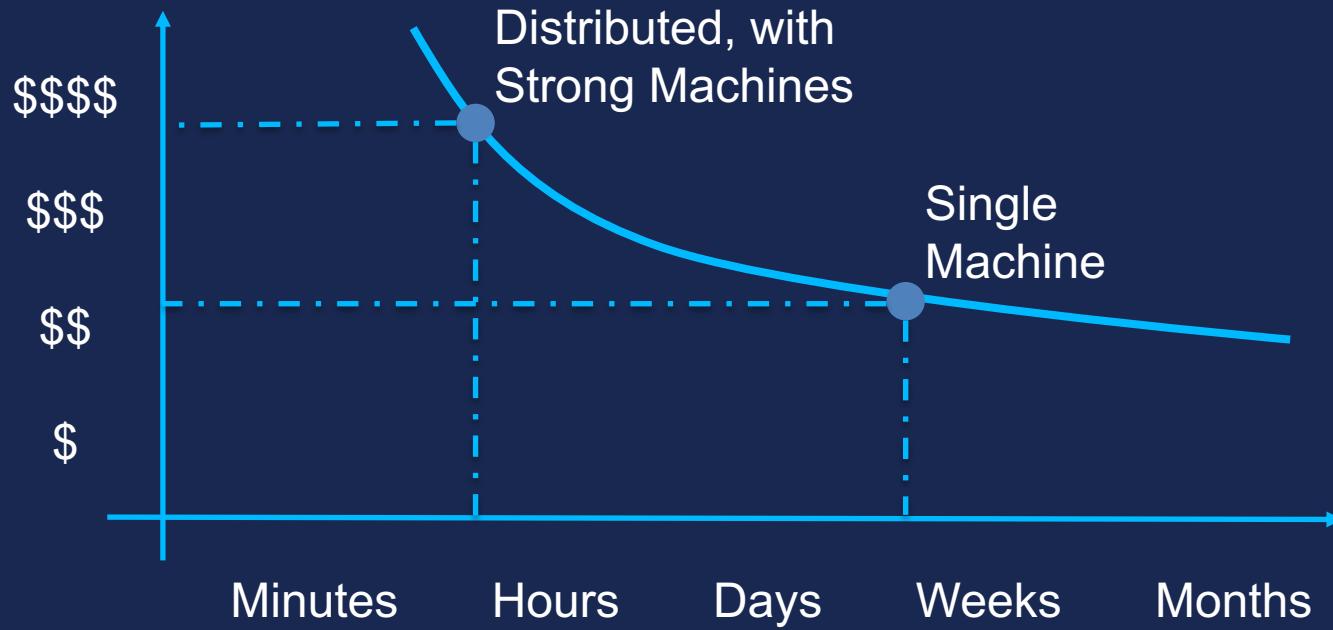
Fully managed
hosting with
auto-scaling

BUILD

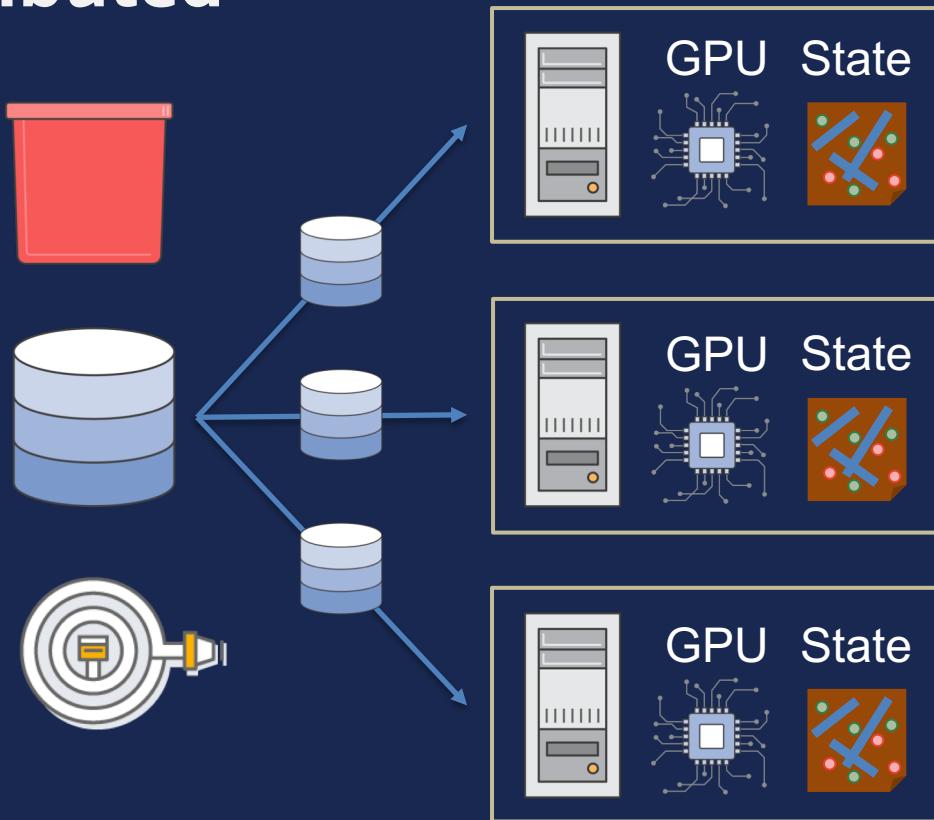
TRAIN

DEPLOY

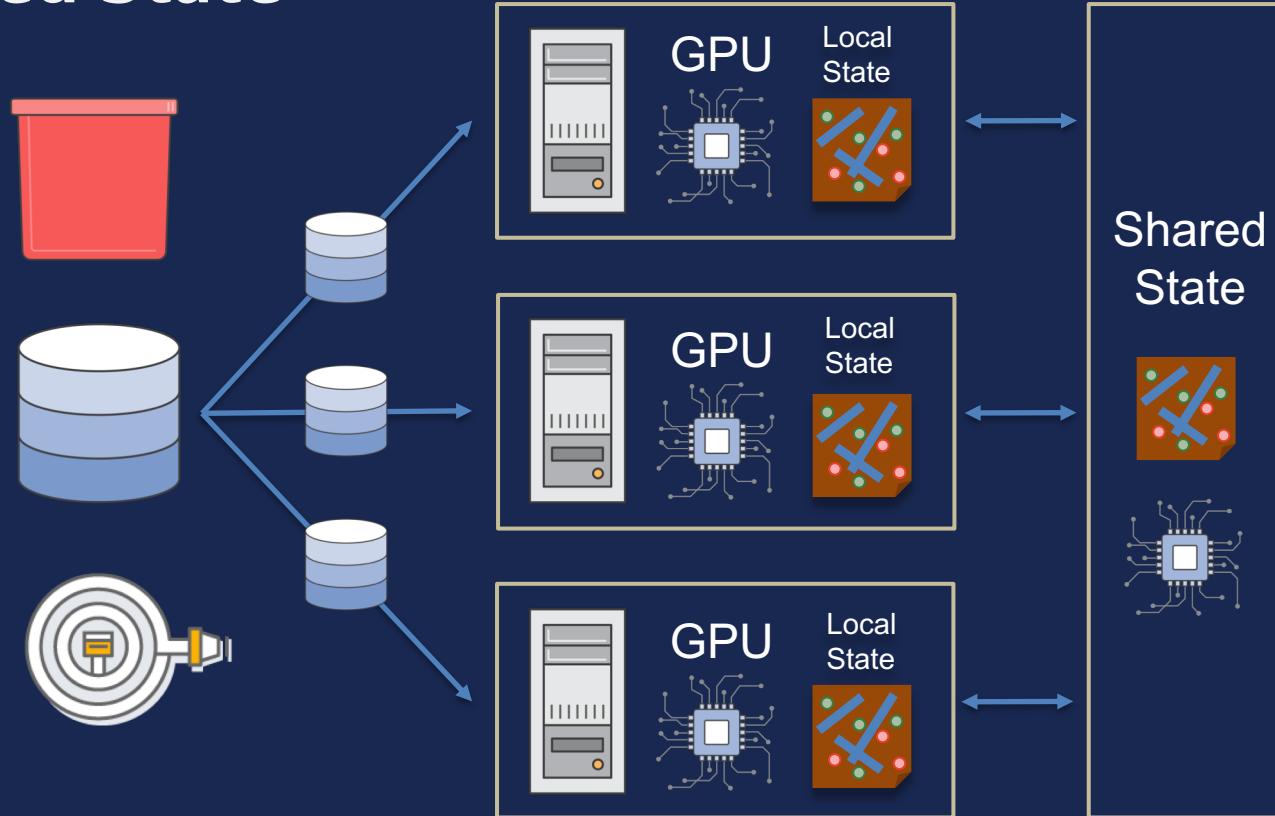
Cost vs. Time



Distributed



Shared State



Cost vs. Time



Recommender Workshop Repository

<http://bit.ly/2wkaV0N>

Our Data Set: MovieLens

- Public Data Set produced by **GroupLens Research**
- <https://grouplens.org/datasets/movielens/>

```
In [15]: data = pd.read_csv("u.data", sep='\t', header=None,
                           names=['userid', 'movieid', 'rating', 'timestamp'])
data.head()
```

Out[15]:

	userid	movieid	rating	timestamp
0	196	242	3	881250949
1	186	302	3	891717742
2	22	377	1	878887116
3	244	51	2	880606923
4	166	346	1	886397596

Recommender Workshop Activity

- Log into the AWS console
- Change to **us-east-1** region
- Find the **Amazon SageMaker** service
- Find **Notebooks**
- Spin up new notebook instance
 - Recommended: m1.m4 type
- Within notebook instance, open Terminal
 - `cd SageMaker`
 - `git clone https://github.com/shirkeyaws/sagemaker-recommender-workshop` (aka `http://bit.ly/2wkaV0N`)
- In Jupyter, within the repo path, find: **01_exploring_data.ipynb**

NEXT: Part 2