

# Amazon SageMaker & Algorithms

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Snr Manager, Specialists Team – AI/ML, Big Data, IoT, Robotics
Amazon Web Services

# Agenda

- Amazon Al Services
- Amazon SageMaker Overview
- Algorithms in SageMaker + demo
- SageMaker Autopilot ("AutoML") + demo
- Using SageMaker for Regression + demo
- SageMaker BYOM, BYOC + demo
- Time series forecasting using Amazon Forecast + demo
- Stretch goal: AWS Step Functions, AWS Lambda, Amazon API Gateway

### How we work with customers

- SAs and Specialist SAs
  - Understand your business needs
  - Strategize on high-impact requirements
  - Deliver workshops and training (use cases, new services)
  - Proof-of-concepts
  - Collaboration on architecture, development, MLOps 🔀



- In addition, AWS offers:
  - Prototyping team (6-week engagements)
  - ML Solutions Lab (for multi-week projects)
  - Professional Services (weeks- to months-long projects)

### **Amazon AI Services**

### The AWS Machine Learning Stack

### Broadest and most complete set of Machine Learning capabilities

#### AI SERVICES



Amazon









**SPEECH** 











Amazon Comprehend Translate

**TEXT** 









Personalize



**FRAUD** 



DEVELOPMENT



Contact Lens For Amazon Connect

**CONTACT CENTERS** 

#### **ML SERVICES**



Ground

**AWS** Marketplace for ML









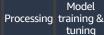












SageMaker Studio IDE



**Autopilot** 

Model hosting

**Model Monitor** 

Neo

Augmented ΑI

#### **ML FRAMEWORKS & INFRASTRUCTURE**







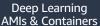












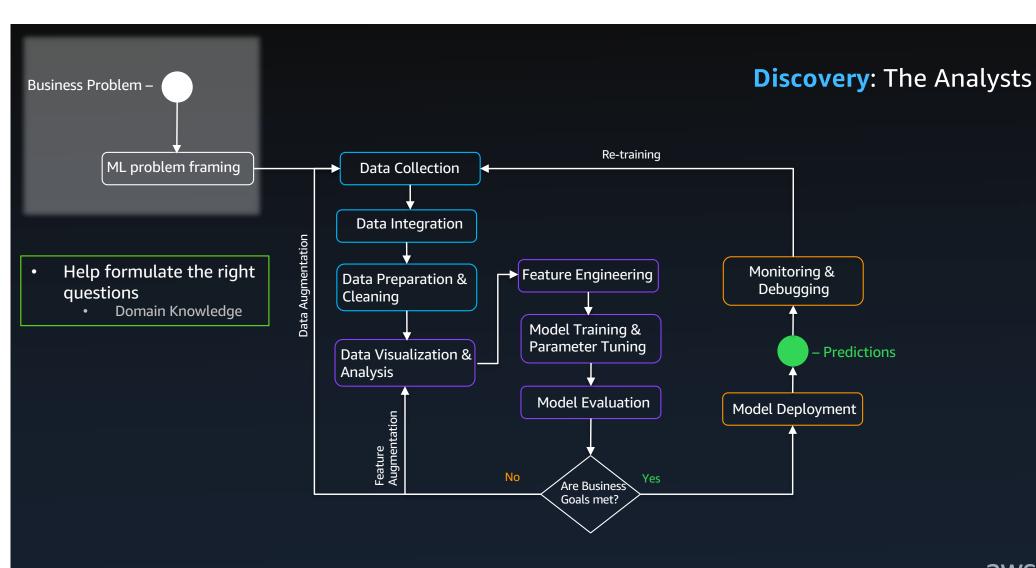


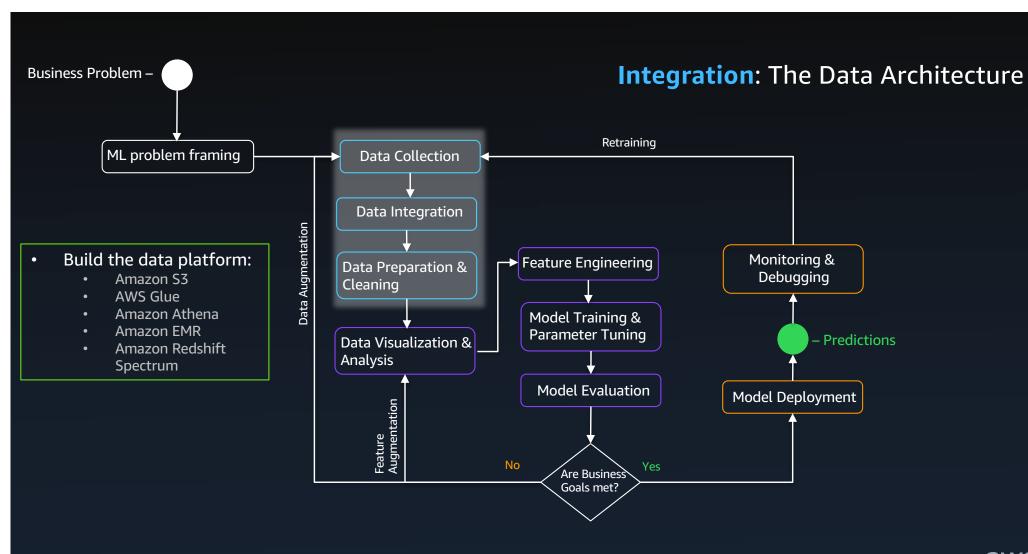
Elastic Inference

Inferentia

**FPGA** 







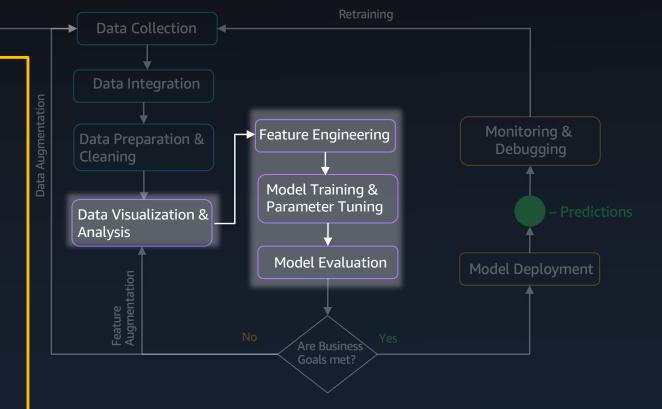
# Business Problem –

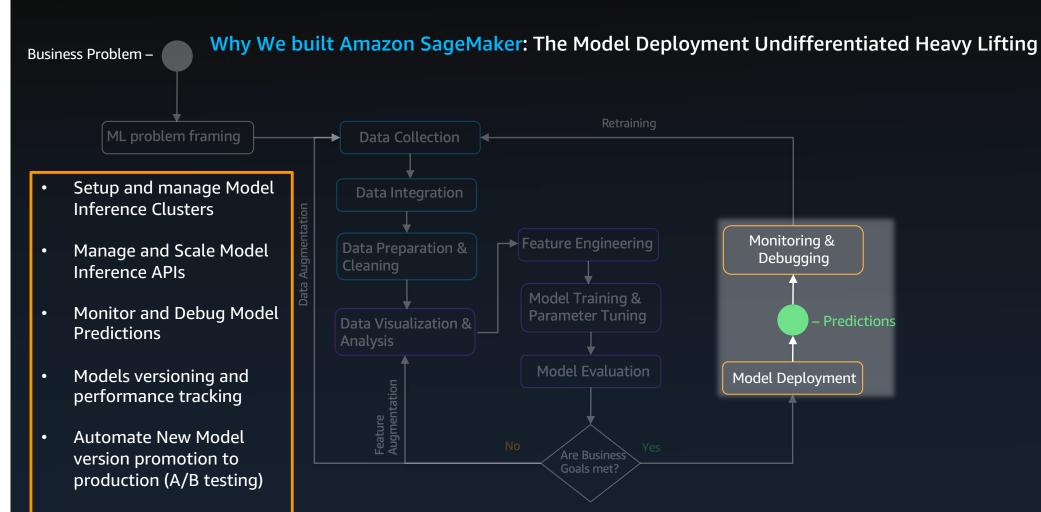
#### Why We built Amazon SageMaker: The Model Training Undifferentiated Heavy Lifting

 Setup and manage Notebook Environments

ML problem framing

- Setup and manage Training Clusters
- Write Data Connectors
- Scale ML algorithms to large datasets
- Distribute ML training algorithm to multiple machines
- Secure Model artifacts





### The AWS Machine Learning Stack

### Broadest and most complete set of Machine Learning capabilities

#### **AI SERVICES**



Amazon

Rekognition









**SPEECH** 





Amazon Comprehend Translate +Medical



TEXT

Amazon Amazon



Amazon Kendra







Personalize



**Forecast** 



**FRAUD** 



DEVELOPMENT



CONTACT CENTERS

#### ML SERVICES



Amazon SageMaker

Ground for ML



Marketplace













#### Experiments







SageMaker Studio IDE



Debugger

Model hosting **Model Monitor** 

Neo

Augmented ΑI

#### **ML FRAMEWORKS & INFRASTRUCTURE**





















Elastic Inference

Inferentia

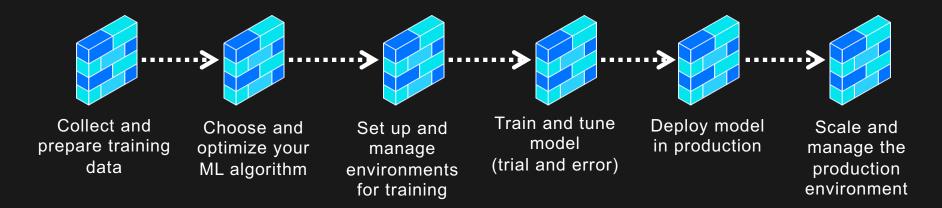
**FPGA** 



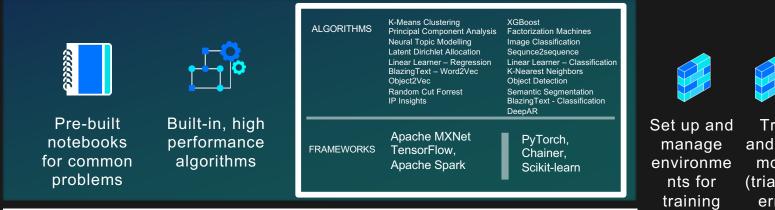


A fully managed service that enables data scientists and developers to quickly and easily build machine-learning based models into production.

Easily build, train, and deploy machine learning models



https://docs.aws.amazon.com/sagemaker/latest/dg/whatis.html https://sagemaker.readthedocs.io/en/stable/ https://docs.aws.amazon.com/sagemaker/latest/dg/r-guide.html









Train and tune model model (trial and error)

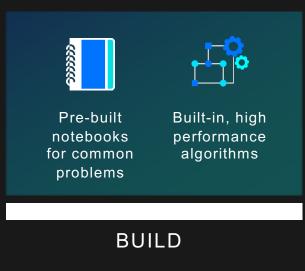
Deploy

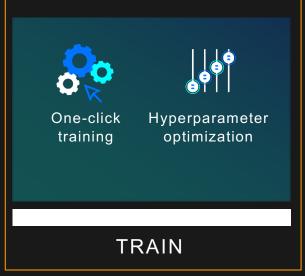
Scale and manage in PROD the PROD environm ent

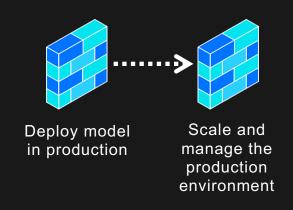
**BUILD** 

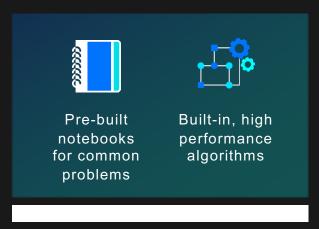
https://docs.aws.amazon.com/sagemaker/latest/dg/frameworks.html

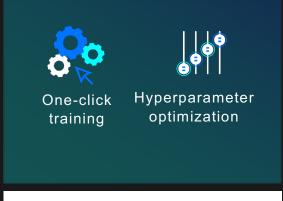








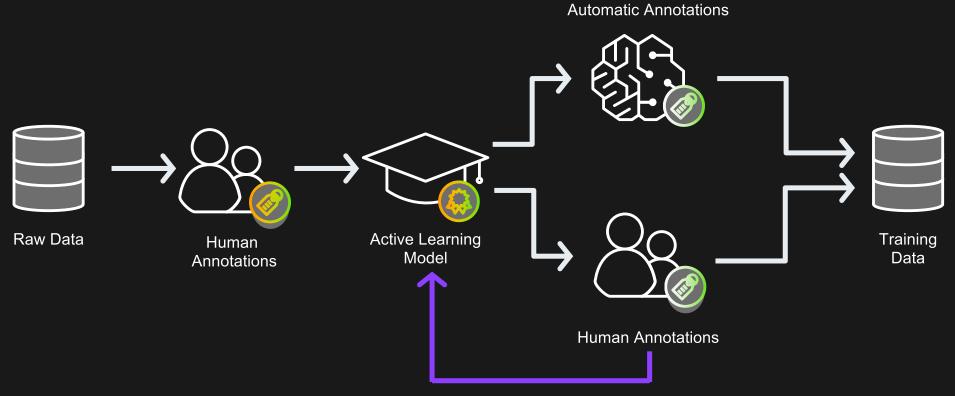






TRAIN BUILD

# Amazon SageMaker Ground Truth: Build highly accurate training datasets and reduce data labeling costs



# Amazon Elastic Inference: Add GPU acceleration to any

Amazon EC2 instance for faster inference at much lower cost



Lower inference costs



Match capacity to demand



Available between 1 to 32 TFLOPS per accelerator

#### KEY FEATURES

Integrated with Amazon EC2 and Amazon SageMaker Support for TensorFlow, Apache MXNet -PyTorch coming soon Single and mixed-precision operations



### **Amazon SageMaker Neo:**

Train once, run anywhere with 2x the performance



Get accuracy and performance



Automatic optimization



Broad framework support



#### KEY FEATURES

Open-source device runtime and compiler, 1/10<sup>th</sup> the size of original frameworks

## **AWS Marketplace for Machine Learning**

ML algorithms and models available instantly



Browse or search **AWS Marketplace** 



Subscribe in a single click



Available in Amazon SageMaker

#### KEY FEATURES

Automatic labeling via machine learning

SELLERS

IP protection

Automated billing and metering

Broad selection of paid, free, and open-source algorithms and models

Data protection

Discoverable on your AWS bill

BUYERS



# Algorithms in SageMaker

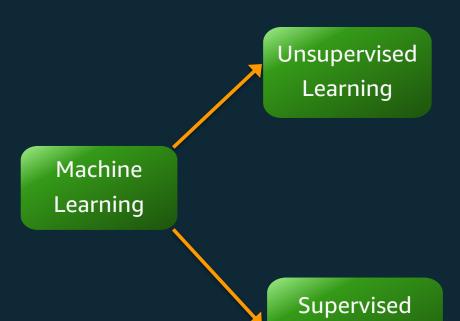
# Common enterprise use cases

Use Case	Approach
Healthcare	develop better processes for diagnosis
Financial Services	prevent fraud, know when to trade, and identify high-risk profiles
Retail	capture, analyze, and use customer shopping data to personalize the shopping experience.
Automotive	improve operations, marketing, and customer experience, as well as quality control vehicle parts.
Government	mine data from multiple sources in order to increase efficiency, save money, detect fraud, and protect against identity theft
Oil & Gas	accurate modeling, optimizing drilling operations, predictive maintenance, subsurface characterization, predicting energy purchasing markets
Manufacturing	automation, quality control, supply chain efficiency, smart factory

### SageMaker Built-in Algorithms

Reinforcement Learning

SageMaker RL



https://docs.aws.amazon.com/sagemaker/latest/dg/algos.html

Learning

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- •K-Means Algorithm
- Principal Component Analysis
- Latent Dirichlet Allocation (LDA)
- Neural Topic Model
- Random Cut Forest
- •IP Insights
- BlazingText\* Word2vec, Object2vec
- Linear Learner
- XGBoost Algorithm
- Factorization Machines
- Image Classification
- Sequence2Sequence
- DeepAR Forecasting
- •K Nearest-Neighbors
- Object detection
- Semantic segmentation
- •BlazingText\* text or document classification

\*Semisupervised



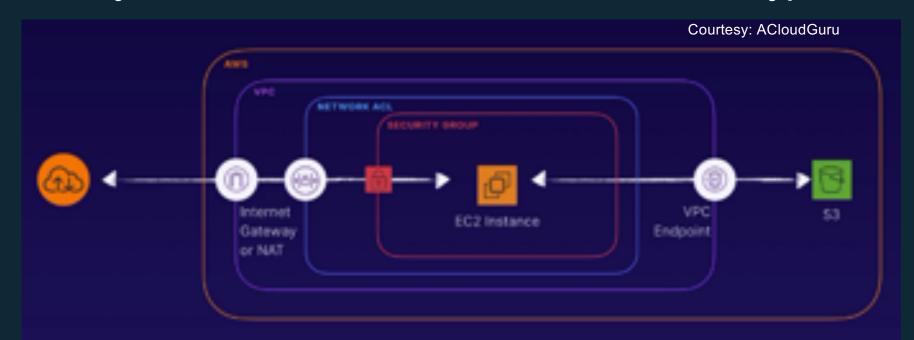
# SageMaker Setup & Console demo

### SageMaker Setup

- Notebook instance https://docs.aws.amazon.com/sagemaker/latest/dg/howitworks-create-ws.html
- IAM role
- S3 bucket
- SageMaker SDK
  - To train, deploy, and validate a model,
    - SageMaker Python SDK or
    - AWS SDK for Python (Boto 3)
  - SageMaker Python SDK abstracts several implementation details, and is easy to use
  - Recommended for first-time users
  - https://sagemaker.readthedocs.io/en/stable/

# SageMaker Security

Visibility, access control, authentication, and encryption



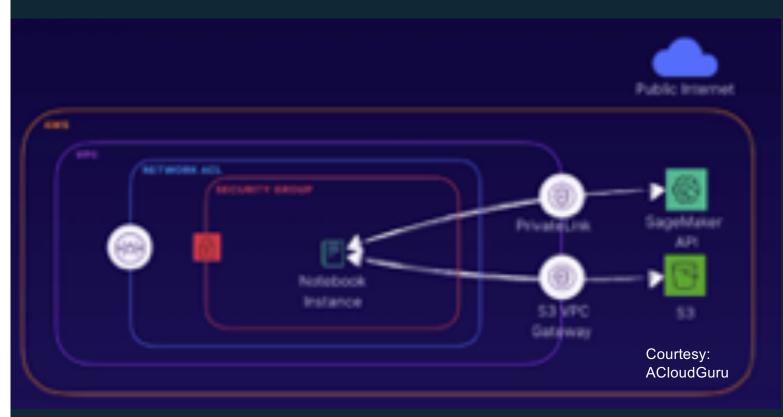
https://docs.aws.amazon.com/sagemaker/latest/dg/security.html https://docs.aws.amazon.com/sagemaker/latest/dg/appendix-notebook-and-

# Notebooks are internet enabled by default



https://docs.aws.a mazon.com/sage maker/latest/dg/m kt-algo-modelinternet-free.html

# With VPC Endpoints ...



- Disable internet access and also permit selected access via NAT GWY, Routes or SGs
- One user/notebook
- Notebooks allow root access
- KMS keys to encrypt data at rest

https://docs.aws.amazon.com/sagemaker/latest/dg/notebook-interfaceendpoint.html

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# Amazon SageMaker Autopilot

Automatic model creation for tabular data with full visibility & control



Quick to start

Provide your data in a tabular form & specify target prediction



Automatic model creation

Get ML models with feature engineering & model tuning automatically done



Visibility & control

Get notebooks for your models with source code

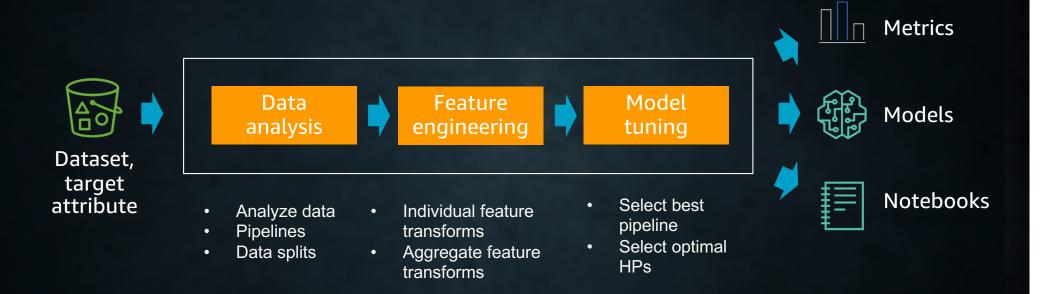


Recommendations & Optimization

Get a leaderboard & continue to improve your model

- Classification, binary and multi-class
- Regression
- https://docs.aws.amazon.com/sagemaker/latest/dg/autopilotautomate-model-development.html

## How SageMaker AutoPilot Works

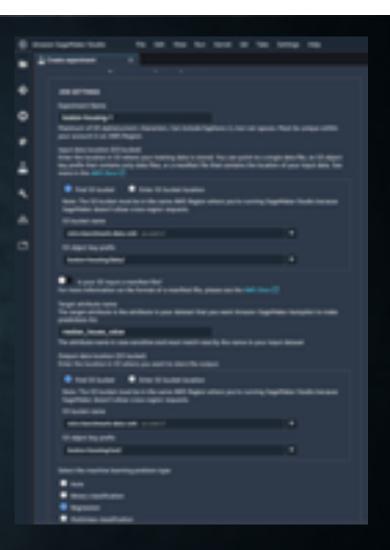


https://github.com/awslabs/amazon-sagemaker-examples/tree/master/autopilot



# SageMaker Autopilot from the Studio

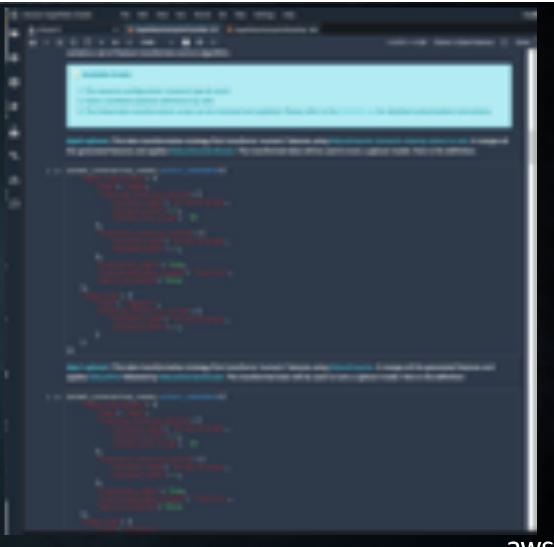
- only S3 location and target variable required
- optional control points:
  - dry-run vs complete mode
  - setting problem type
  - security settings
- API level control points:
  - number of candidate models to build
  - maximum time to take
  - model evaluation metric (accuracy, F1, RMSE)



# Autopilot

Fully runnable model candidate notebook:

- data transformers
- featurization techniques applied
- override points:
  - · algorithms considered
  - evaluation metric
  - hyper-parameter ranges
  - model search strategy
  - instances used



## What Autopilot does well

- AutoML for classification and regression learning
- Exhibits model transparency and extensibility
  - White box approach
- Data analysis → data properties → feature engineering candidate generation → multiple candidate pipelines
- XGBoost and linear-learner algorithms
  - Scalable, can run distributed for large datasets (5GB)
  - More algorithms to come
  - Up to 10 different candidate pipelines are run in parallel
- Handles both numerical and text data (will featurize text with TF-IDF, etc.

# SageMaker Autopilot

# Demo

# **Specific Use Cases**

- Regression
  - Two options
    - Autopilot
    - Built-in SageMaker algorithms
- Time series forecasting
  - Two options
    - Amazon Forecast
    - Built-in SageMaker algorithm

# How to use SageMaker Algorithms for Regression

- Multiple SageMaker algorithms, e.g., Linear Learner and XGBoost
  - For regression (sales forecasting, predicting delivery times)
    - Set hyperparameter predictor\_type = regressor (for linear-learner)
    - Set hyperparameter objective = reg:linear (for XGBoost)
  - For classification (ad-click prediction, customer churn)
    - Set hyperparameter predictor\_type = binary\_classifier (for linear-learner)
    - Set hyperparameter objective = reg:logistic (for XGBoost)
  - Ensembling, aka using both

## Regression Example

- **Using XGBoost**
- https://github.com/awslabs/amazon-sagemakerexamples/blob/master/introduction\_to\_amazon\_algorithms /xgboost\_abalone/xgboost\_abalone.ipynb

Console demo

# Model Deployment – BYO Algorithms or Models

- SageMaker uses Docker containers for build and runtime tasks
- Put scripts, algorithms, and inference code of your MLmodels into containers
- Package your training code, inference code
- Four options:
  - 1. Use a built-in algorithm
  - 2. Use pre-built container images that supports Deep Learning frameworks
  - 3. Extend a pre-built container image (e.g., PyTorch)
  - Build your own custom container image

https://docs.aws.amazon.com/sagemaker/latest/dg/your-algorithms.html

# Bring-Your-Own-Code Inferencing (BYOM)

- Train your own model
- Model file name must satisfy this RE pattern
- Model file has to be tar-zipped
- Upload your model to S3
- Import model into hosting (scikit-learn XGBoost model is compatible with SageMaker XGBoost container, other gradient boosted tree models are not)
- Create end-point configuration with model name (now in S3)
- Create end-point
- Run inferencing
- https://docs.aws.amazon.com/sagemaker/latest/dg/your-algorithmsinference-code.html
- Example notebooks: <u>BYOM XGBoost</u>, <u>BYOM K-Means</u>



# Bring-Your-Own-Code Training and Hosting

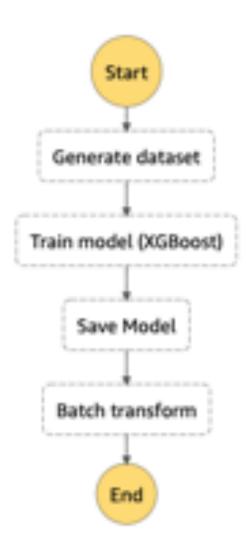
- Package your own algorithm for training and deployment
- Bring any code to SageMaker regardless of programming language, environment, framework, etc.
- Why build your own container?
  - Complex algorithm
  - Special additions to framework
- No need to provide your container for common frameworks
  - Provide code that implements your algorithm
- Add additional permissions: *AmazonEC2ContainerRegistryFullAccess*
- Build the image files (Docker)
- One Docker image for training and hosting or two separate
- How to: <a href="https://github.com/aws/sagemaker-training-toolkit">https://github.com/aws/sagemaker-training-toolkit</a>
- Example notebooks: <u>BYOM Scikit</u>, <u>BYO R</u>, <u>BYO Host Multiple Models</u>, <u>BYOC TF</u>

# Using your custom algorithm in SageMaker

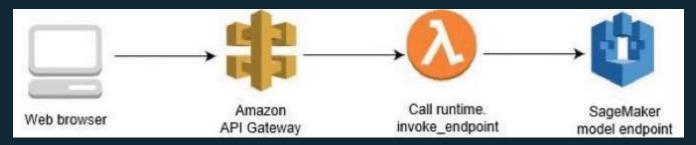
- Define Docker image as described earlier
- Register with SageMaker image registry (ECR)
- Create code entry points as described earlier
- Pass image to SageMaker estimator function
- Fit the model
- Deploy the model for real-time prediction or Batch
- Run inference
- BYOC TF

# Using SageMaker with AWS Step Functions

- Using AWS Step Functions to manage batch training:
- https://docs.aws.amazon.com/stepfunctions/latest/dg/sample-train-model.html
- Notebook: <a href="https://github.com/juliensimon/amazon-sagemaker-examples/blob/master/step-functions-data-science-sdk/machine learning workflow abalone/machine learning workflow w abalone.ipynb">https://github.com/juliensimon/amazon-sagemaker-examples/blob/master/step-functions-data-science-sdk/machine learning workflow abalone.ipynb</a>
- https://www.youtube.com/watch?v=0kMdOi69tjQ



### Calling SageMaker Endpoints using Amazon API Gateway



- How do we use a hosted SageMaker model?
- Create a SageMaker endpoint → Call using SageMaker run-time API
  - You need infrastructure to host that invocation code
- Can we make this independent of infrastructure? Yes
- Use Lambda to invoke that endpoint (SageMaker API is embedded as Lambda function): <a href="https://docs.aws.amazon.com/lambda/latest/dg/welcome.html">https://docs.aws.amazon.com/lambda/latest/dg/welcome.html</a>
- Call Lambda from an API Gateway
   (https://docs.aws.amazon.com/apigateway/latest/developerguide/welcome.html)

# Additional Workbooks to try and **Documentation Links**

### Other useful links: AWS development

- SageMaker Python SDK: <u>https://github.com/aws/sagemaker-python-sdk-overview</u>
- AWS Python SDK: <a href="https://aws.amazon.com/sdk-for-python/">https://aws.amazon.com/sdk-for-python/</a>
- Boto3 for SageMaker: <u>https://boto3.amazonaws.com/v1/documentation/api</u> /latest/reference/services/sagemaker.html

### SageMaker Notebooks Doclinks

- Create an S3 bucket: https://docs.aws.amazon.com/sagemaker/latest/dg/gs-configpermissions.html
- Create a SageMaker notebook instance: https://docs.aws.amazon.com/sagemaker/latest/dg/gs-setupworking-env.html
- Customize a notebook instance (optional): https://docs.aws.amazon.com/sagemaker/latest/dg/notebooklifecycle-config.html
- Additional exercises (for homework) https://docs.aws.amazon.com/sagemaker/latest/dg/ex1.html

## SageMaker Operations - Doclinks

- 1. Monitor and visualize: <a href="https://aws.amazon.com/blogs/machine-learning/easily-monitor-and-visualize-metrics-while-training-models-on-amazon-sagemaker/">https://aws.amazon.com/blogs/machine-learning/easily-monitor-and-visualize-metrics-while-training-models-on-amazon-sagemaker/</a>
- 2. Using common workflows for cloud-based development: <a href="https://aws.amazon.com/blogs/machine-learning/how-to-use-common-workflows-on-amazon-sagemaker-notebook-instances/">https://aws.amazon.com/blogs/machine-learning/how-to-use-common-workflows-on-amazon-sagemaker-notebook-instances/</a>
- Invoke the model as an endpoint using API Gateway and Lambda: <a href="https://aws.amazon.com/blogs/machine-learning/call-an-amazon-sagemaker-model-endpoint-using-amazon-api-gateway-and-aws-lambda/">https://aws.amazon.com/blogs/machine-learning/call-an-amazon-sagemaker-model-endpoint-using-amazon-api-gateway-and-aws-lambda/