Data Science in the Cloud

With 3 approaches



Motivation

Data science and ML often require a **lot of compute resources** that developers' PCs might not have.

Jupyter notebooks on Kaggle or Colab are either **limited or not highly available**, so lets run jupyter servers on the cloud and work there:

- 1. Approach: Manually create EC2 and install conda & jupyter lab
- 2. Approach: Automate and expand setup of 1st approach
- 3. Approach: Use Sagemaker instance notebooks or studio

The AWS ML Stack

Broadest and most complete set of machine learning capabilities

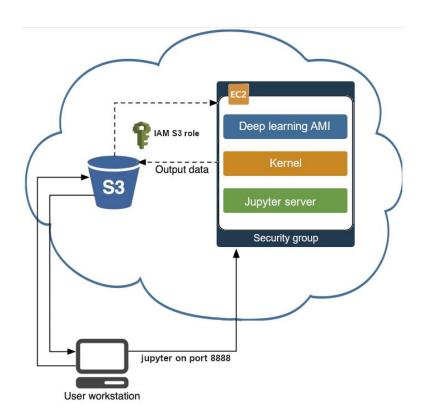




Steps to run a jupyter notebook on an EC2

- 1. Expose the port 8888 in security group (limit to your IP)
- 2. Get and install conda
- 3. Install all required packages
- 4. Configure the jupyter configuration (add encrypted password)
- Initialize conda
- 6. Restart bash and activate conda environment
- 7. Start jupyter notebook
- 8. Access jupyter notebook on public IP:8888 of EC2

Automation with Terraform



Main.tf

- Specify all infrastructure needed
- Auto Scaling group
- AMI
- Security group
- IAM S3 role and policy

Variables.tf

Secret keys, instance types etc.

Install.sh

Script to set python env and start jupyter





Prepare \rightarrow

SageMaker Ground Truth
Label training data for machine
learning

SageMaker Data Wrangler NEW Aggregate and prepare data for machine learning

SageMaker Processing Built-in Python, BYO R/Spark

SageMaker Feature Store NEW Store, update, retrieve, and share features

SageMaker Clarify NEW
Detect bias and understand
model predictions

Build \rightarrow

SageMaker Studio Notebooks
Jupyter notebooks with elastic
compute and sharing

Built-in and Bring-your-own AlgorithmsDozens of optimized algorithms or bring
your own

Local Mode
Test and prototype on your local machine

SageMaker Autopilot
Automatically create machine
learning models with full visibility

SageMaker JumpStart NEW
Pre-built solutions for common
use cases

Train & tune \longrightarrow

One-click Training
Distributed infrastructure
management

SageMaker Experiments
Capture, organize, and compare
every step

Automatic Model Tuning Hyperparameter optimization

Distributed Training Libraries NEWTraining for large datasets
and models

SageMaker Debugger NEW Debug and profile training runs

Managed Spot Training Reduce training cost by 90%

Deploy & manage →

One-click Deployment
Fully managed, ultra low latency,
high throughput

Kubernetes & Kubeflow IntegrationSimplify Kubernetes-based machine learning

Multi-Model Endpoints
Reduce cost by hosting multiple
models per instance

SageMaker Model Monitor Maintain accuracy of deployed models

SageMaker Edge Manager NEW Manage and monitor models on edge devices

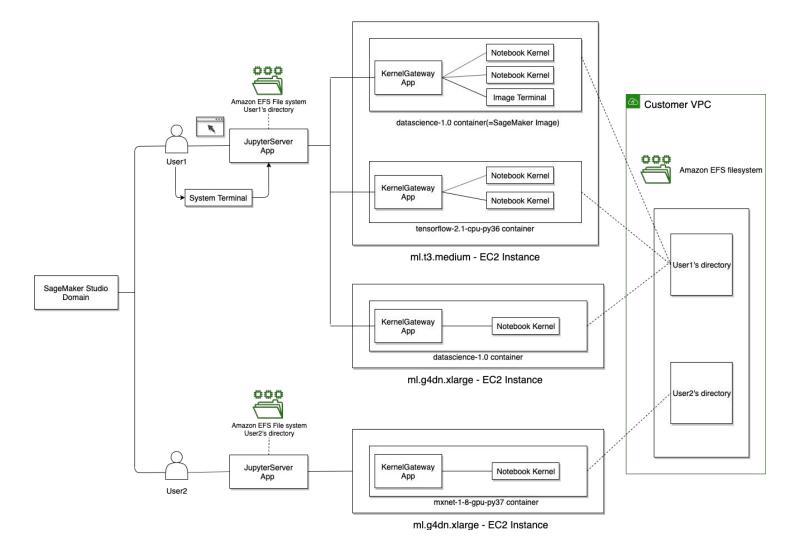
SageMaker Pipelines NEW Workflow orchestration and automation

SageMaker Studio

Integrated development environment (IDE) for ML



Live Demo
Amazon SageMaker Studio



Conclusions & Future work

Sagemaker currently offers many tools for data science IDEs:

- Many capabilities to get anyone started with ML
- Perfect from separating data scientists from cloud computing
- Good tool for implementing MLOps methods

In the future:

- Work on my automation script to ensure data and model persistence across sessions
- Emulate the Studio setup to improve collaboration



Thank you! Any questions?