Operationalizing an AWS ML Project

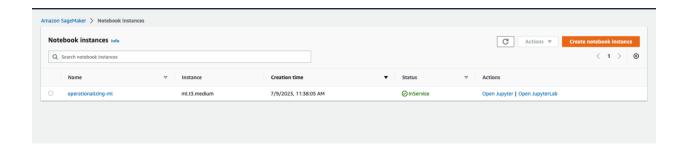
1. Initial Setup

I have chosen the "ml.t3.medium" instance type for this Notebook due to a variety of considerations, all of which cater to the specific needs of this project.

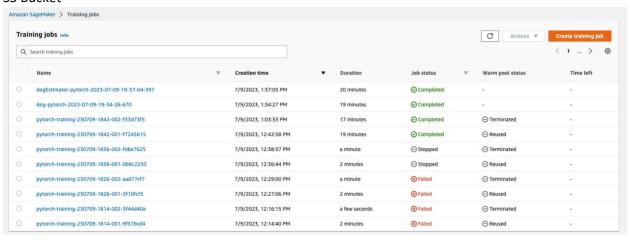
Firstly, the nature of our project doesn't necessitate the employment of a high-performance CPU or an abundance of RAM for the successful execution of our Jupyter notebooks. Rather than focusing on raw computational power, it's more prudent to consider the time duration our notebook instance will remain active.

Considering the potentially substantial duration of our project, it's essential to select an instance with a cost-effective hourly rate, in addition to providing a reasonable level of CPU and RAM. This strategic decision is crucial to curbing excessive costs while ensuring the functionality and efficiency of our project are uncompromised.

In light of these considerations, the "ml.t3.medium" instance emerges as an optimal choice. It provides a balance between cost-efficiency and computational prowess that fits our project's requirements. Despite a slower startup time, this instance type is more cost-effective per hour, a trade-off that is acceptable given our project's lack of dependency on instant start-up.



S3 Bucket



Training and Tuning Jobs

Objects (3)

C Copy S3 URI

Q Find objects by prefix

test/

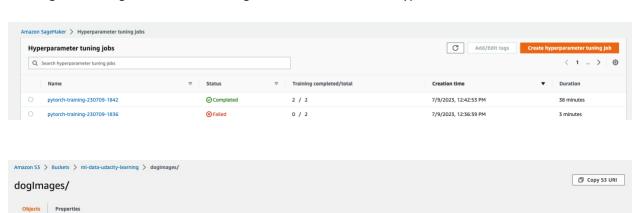
□ valid/

Training and Tuning didn't take that long because of the instance type I used.

▼ Last modified

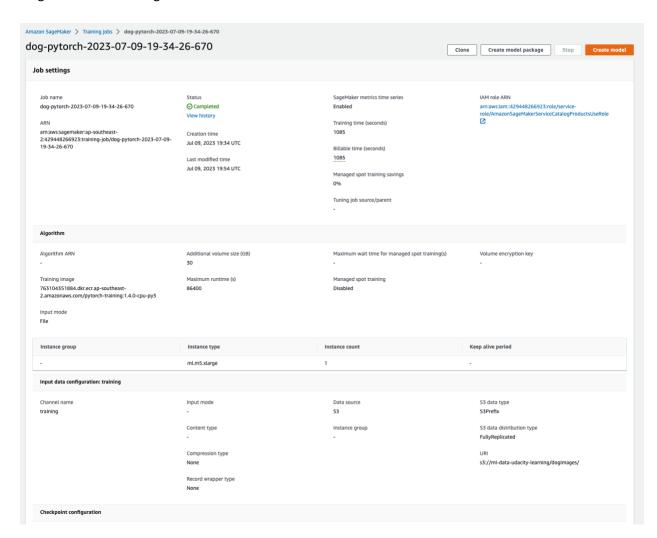
▲ Type

Folder



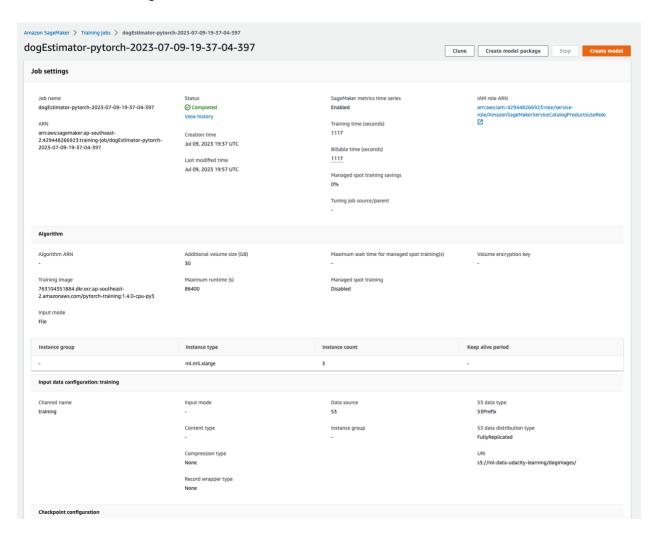
▽ Storage class

Single Instance Training Job

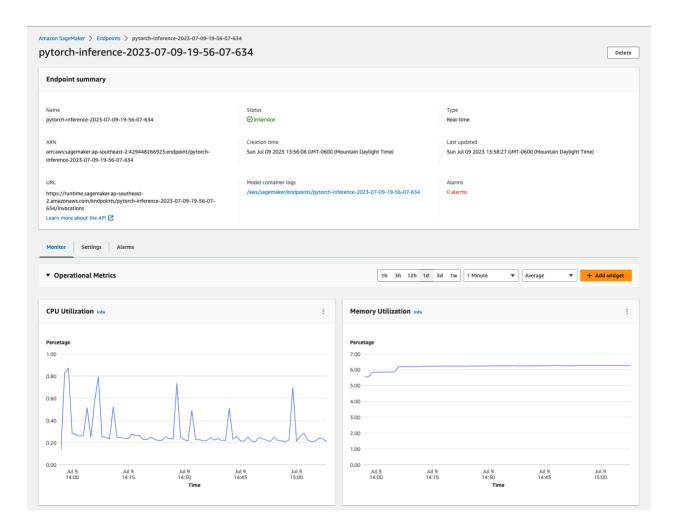


Both Training Jobs took fairly equal amount of time and I am very surprised.

Multi-Instance Training Job:



Deployment Endpoints



2. EC2 Training

I employed the Deep Learning AMI, specifically GPU PyTorch 2.0.1 (Amazon Linux 2) 20230627, ami-051619310404cab17 (64-bit (x86)), alongside the g4dn.xlarge instance. This combination strikes an effective balance between financial considerations and performance capabilities.

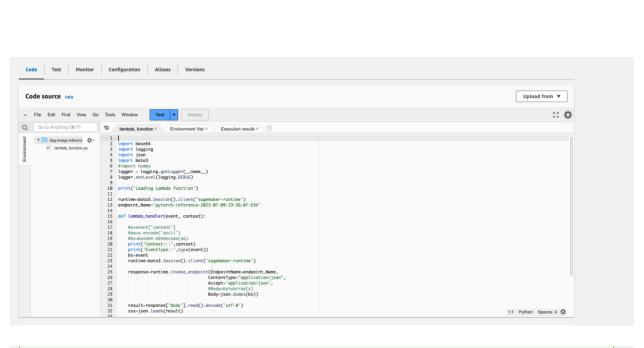
To begin with, g4dn.xlarge instances emerge as the most economical option for this project, primarily due to the fact that the g4dn series generally represents the least expensive instances compatible with the Deep Learning AMI.

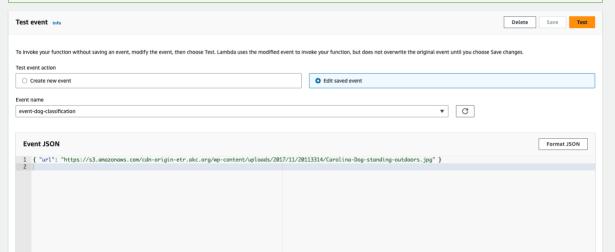
According to the official documentation, g4dn.xlarge instances are capable of maintaining superior CPU performance for the duration required by a task, a characteristic that adds to their suitability for this project.

Lastly, without incurring any additional expenses, g4dn instances provide adequate performance for the majority of applications of a general nature.

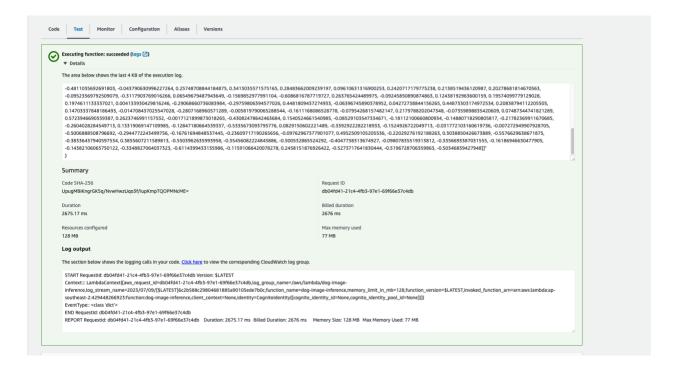
This instance type is supported by several EC2 instances, namely G3, P3, P3dn, P4d, P4de, G5, and G4dn. For further details, you can refer to the release notes available at: https://docs.aws.amazon.com/dlami/latest/devguide/appendix-ami-release-notes.html.

```
🧿 🔵 🥛 Townloads — ec2-user@ip-172-31-89-231:~ — ssh -i udacityLearn.pem ec2-user@ec2-54-85-187-205.compute-1.amazonaws.com — 148×62
nltk_data
node_modules
opt
package-lock.json
package.json
postgresql_14.app.zip
pyconfig.h
pycomig.n
serverless.yml
tmp
(base) sulavdahal@Sulavs-MacBook-Pro - % cd Downloads
(base) sulavdahal@Sulavs-MacBook-Pro Downloads % ssh -i "udacityLearn.pem" ec2-user@ec2-54-85-187-205.compute-1.amazonaws.com
  e) sult.
se) sultv.
, #_
, "###
- \####|
\###|
\##-
\/-'->
                                        Amazon Linux 2023
                                       https://aws.amazon.com/linux/amazon-linux-2023
           _/m/'
Last login: Sun Jul 9 20:14:33 2023 from 192.225.179.230
[ec2-user@ip-172-31-89-231 ~]$ ls
TrainedModels doginages.zip solution.py
[[ec2-user@ip-172-31-89-231 ~]$ cat solution.py
import numpy as np
import torch
import torch.nn as nn
import torch.optim as optim
import torchvision
import torchvision.models as models
import torchvision.transforms as transforms
import copy
import argparse
import os
import logging
import sys
from tydm import tqdm
from PIL import ImageFile
ImageFile.LOAD_TRUNCATED_IMAGES = True
#rom torch_snippets import Report
#from torch_snippets import *
logger=logging.getLogger(__name__)
logger.setLevel(logging.DEBUG)
logger.addHandler(logging.StreamHandler(sys.stdout))
def test(model, test_loader, criterion):
    model.eval()
       running_loss=0
running_corrects=0
       for inputs, labels in test_loader:
   outputs=model(inputs)
   loss=criterion(outputs, labels)
   _, preds = torch.max(outputs, 1)
   running_loss += loss.item() * inputs.size(0)
   running_corrects += torch.sum(preds == labels.data)
```





A SageMaker Policy has been attached to make the lambda code functional.



Concurrency

