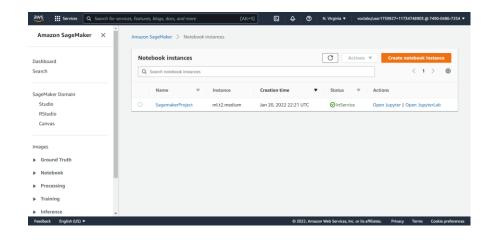
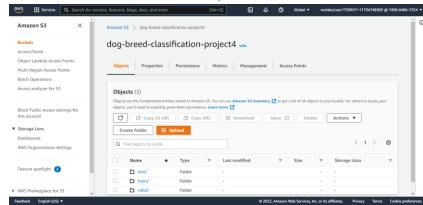
## Operationalizing Machine Learning on SageMaker

# Training and Deployment 1

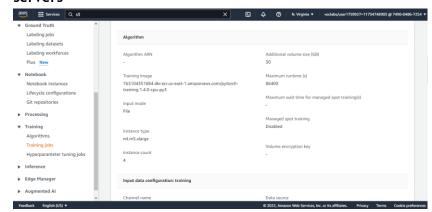
- Every notebook in SageMaker needs a computing instance in order to run and we use the notebook instance to create and manage Jupyter notebooks for preprocessing data and to train and deploy machine learning models.
- There are many types of instances, and every time we run an instance we will need to make a good choice for the type of instance to run.
- The instance type we choose needs to be sufficient for your computing needs, but it should also minimize costs.
- Though, I have chosen the "ml.t2.medium" instance type for Notebook instance.
- And to guarantee for completing the execution of this particular project's jupyter notebooks we do not need a very computationally powerful CPU and high RAM.
- But We will need to keep this notebook instance in "in Service" status for a long time while we are working on the project So, avoid high costs we should select a notebook that is low in per hour cost and also offers reasonably good CPU and RAM.



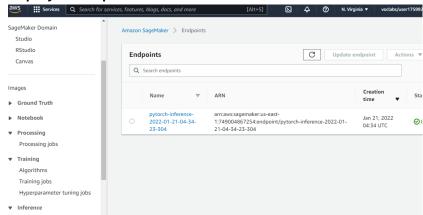
 The dog breed dataset for this was uploaded to a newly created S3 bucket, successfully.



- At hyperparameter Tuning I used
  - Instance type= ml.p3.2xlarge
  - Max jobs = 2
  - Max parallel jobs = 2
- Multi-instance Training
- training data on multiple machines is very hard to implement and requires a lot of intricate configurations to get different machines to work together.
- One of the great things about AWS and SageMaker is that they make multi-instance training much easier.
- Multi-instance training in a term for fitting is machine learning models using multiple separate computers or servers



#### Deployed Endpoints:



- Using a special type of EC2 instance called a Spot Instance minimize costs is very important.
- EC2 Instances: computing resources that are less expensive than SageMaker instances, but offer fewer managed services
- Spot Instances: special types of EC2 instances that consist of idle resources that other AWS customers have reserved. They are less expensive than EC2 instances, but they are not reliable in general.
- During the process of opening a new instance, I was able to select an option to launch a spot instance instead of a standard EC2 instance.
- And At Launch Instances I needed to select what's called an AMI, Amazon Machine Image and this is essentially for the operating system.
- During the process of opening a new instance I faced no problems lunching:



 Then I can choose an instance type choice and because the model training could take long, I had to change the default instance type to:



- To ensure Flawless model training.
- The important step is where I configure the instance in order to make it a Spot Instance.
- Though, the costs for Spot Instances are usually quite low because these are idle extra resources.
- And to prepare for EC2 model training I Start by launching and connecting to an EC2 instance as described in the project.
- But when I started the Training, I actually faced multiple errors which was Unable to activate multiple environments.
- And with Udacity mentor help he advised me to change the AMI to be:



- As the required environments pre-installed.
- And Finally, the Model Training was done in Less time and flawlessly aslo saved.
- ★ Using Amazon EC2 eliminates our need to invest in hardware up front, so you can develop and deploy applications faster.
- ★ Also, we can use Amazon EC2 to launch as many or as few virtual servers as we need, configure security and networking, and manage storage.
- **★** Also, Amazon EC2 provides Virtual computing environments, known as instances.

- ★ Also, Various configurations of CPU, memory, storage, and networking capacity for our instances, known as instance types.
- ★ Last but not Least, there are many other Amazon EC2 features provided for us.

## Lambda Function

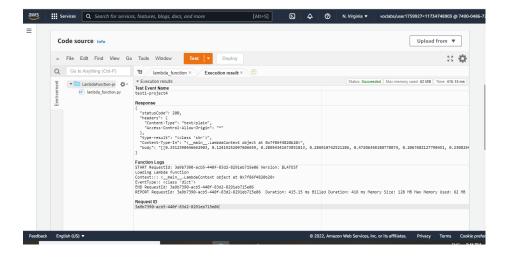
- Lambda functions enable my model and its inferences to be accessed by API's and other programs, so it's a crucial part of production deployment.
- I have sated up a Lambda function that uses Python 3 for its runtime.
- And I attached in the Lambda function the starter code provided by the project resources.
- in this file called lambdafunction.py contains some of the basic Python code for a Lambda function, but I needed to make an important adjustment which was changing the endpoint name endpoint\_name variable, to give it the same name as the endpoint I deployed in the first in the project.

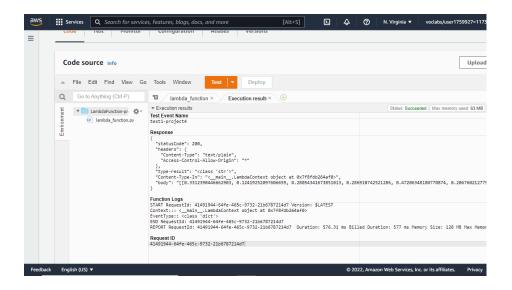
### Lambda function security

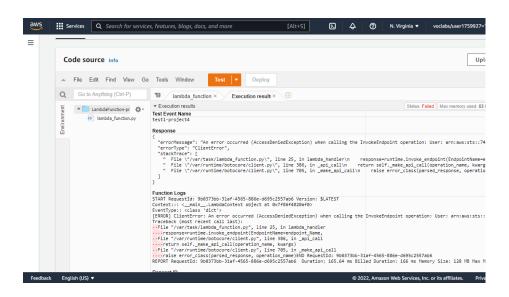
4

- By default, Lambda functions are not given permission to invoke your SageMaker endpoints.
- And In order to allow the Lambda function to invoke my endpoint, I need to make adjustments to the IAM settings.
- Making security adjustments like this is common in industrial ML projects as mentioned in the lesson.
- When I tested my Lambda function:
  - an error that says that this particular role is not authorized to invoke the endpoint.

- This is a default security setting to keep my projects safe.
- And to solve this issue I need to navigate to IAM:
  - Then, in "Roles" section in the role associated with my Lambda function.
  - Inside this role, the policies that are attached to my role.
  - So, I attached a policy called AmazonSageMakerFullAccess.
  - This is a policy that will allow my Lambda function to access any resource in SageMaker.
- Finally, Lambda function, and succussed on the test event.
- In result, Lambda function is delivering model predictions without any security.
- And yes, I do believe that AWS workspace is secure and by using Amazon Workspaces, we have no data leakage and no copying or downloading of files and data.
- Also, AWS has improved flexibility, scalability, and reliability on the Workspaces.
- With AWS Workspaces we can securely access various data sources and we can control who we give access to and what tools and level of access each user gets.
- We have recognized benefits that the AWS Workspaces technology provides such as flexibility, scalability and secure access.







#### **S** Concurrency

- Concurrency refers to the ability of a Lambda function to serve multiple requests simultaneously.
- I set up concurrency for my Lambda functions to allow our Lambda function to be able to access three instances to reply to multiple requests simultaneously.
- This is called reserved concurrency and The advantage is that reserved concurrency has a relatively low cost.
- The other type of concurrency is called provisioned concurrency.
- Provisioned concurrency creates instances that are always on and can reply to all traffic without requiring a wait for startup times.
- Though it can be more flexible and it can achieve low latency even in very high traffic scenarios.
- Choosing a high number for provisioned concurrency is suitable for very high-traffic projects, because it will turn on instances that can be used by your Lambda function any time.
- Choosing a high number for reserved concurrency incurs no additional cost so in this project I have the flexibility to choose any number in this project was 5.
- It will allow me to choose a high number for provisioned concurrency, since provisioned concurrency must be lower than reserved concurrency.
- Choosing lower numbers for either or both types of concurrency would be more suitable for lower-traffic projects, or projects with lower budgets.
- provisioned concurrency is expensive, so I kept this number low which was 2.

#### Auto-Scaling 6

- Autoscaling for endpoints allows my deployed endpoints to respond to multiple requests at the same time.
- So, I Choose my maximum instance count to be = 3
  - My endpoint will be able to automatically scale to any number of instances up to the maximum instance count I select.
- Then I Choose what's called a scale-in cool down time period to be = 30.
  - The scale-in period is the amount of time AWS will wait before deploying more instances for your endpoint.
  - If I choose a high number, then AWS will wait a longer time before deploying more instances.
  - And, this helps me avoid incurring more costs for momentary spikes in traffic.
  - If I choose a low number, then AWS will deploy instances more quickly, but this responsiveness will be more costly.
- Choose a scale-out cool down time period also to be =30.
  - The scale-out cool down period is the amount of time AWS will wait before deleting extra deployed instances.
  - If I choose a low number, then AWS will wait only a short time before deleting extra deployed instances.
  - And, this helps you avoid incurring costs for momentary spikes in traffic.
  - However, If I choose a high number, then AWS will keep extra instances deployed longer, but this extra capacity will be more costly.