Lab Assignment-11

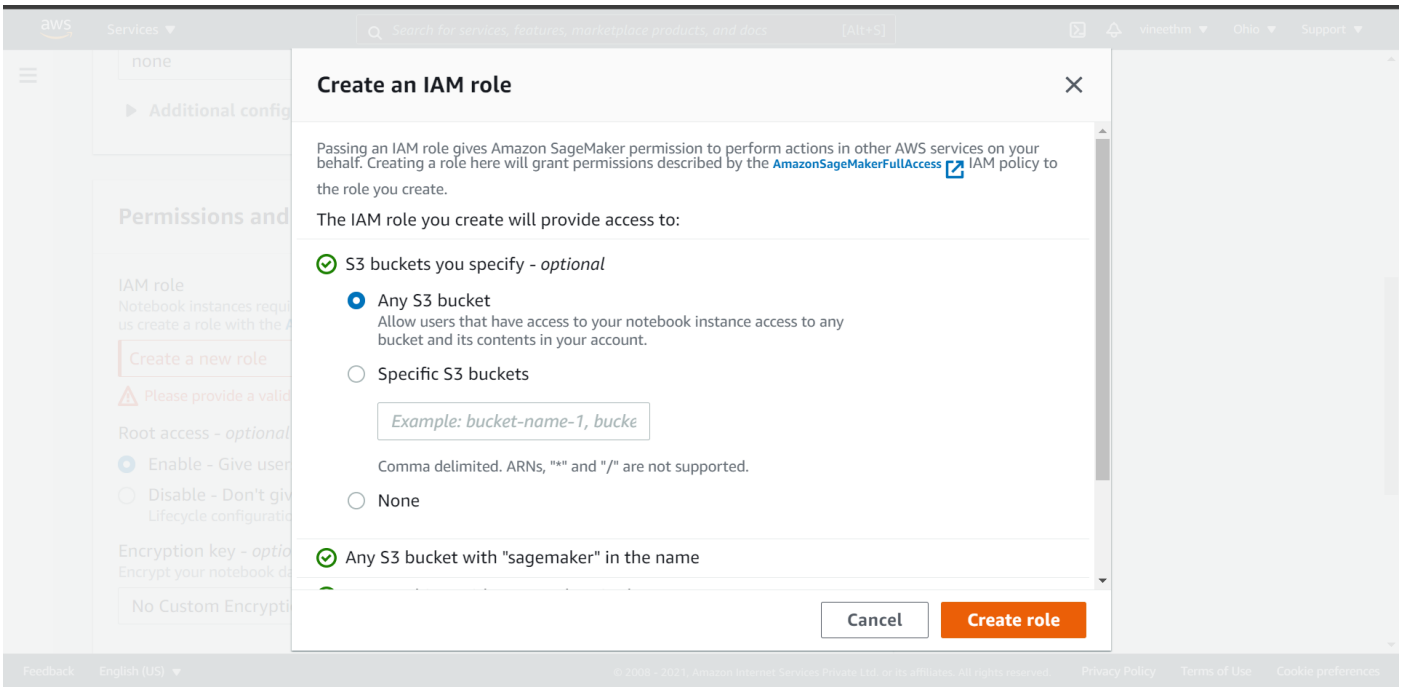
Working with Amazon SageMaker

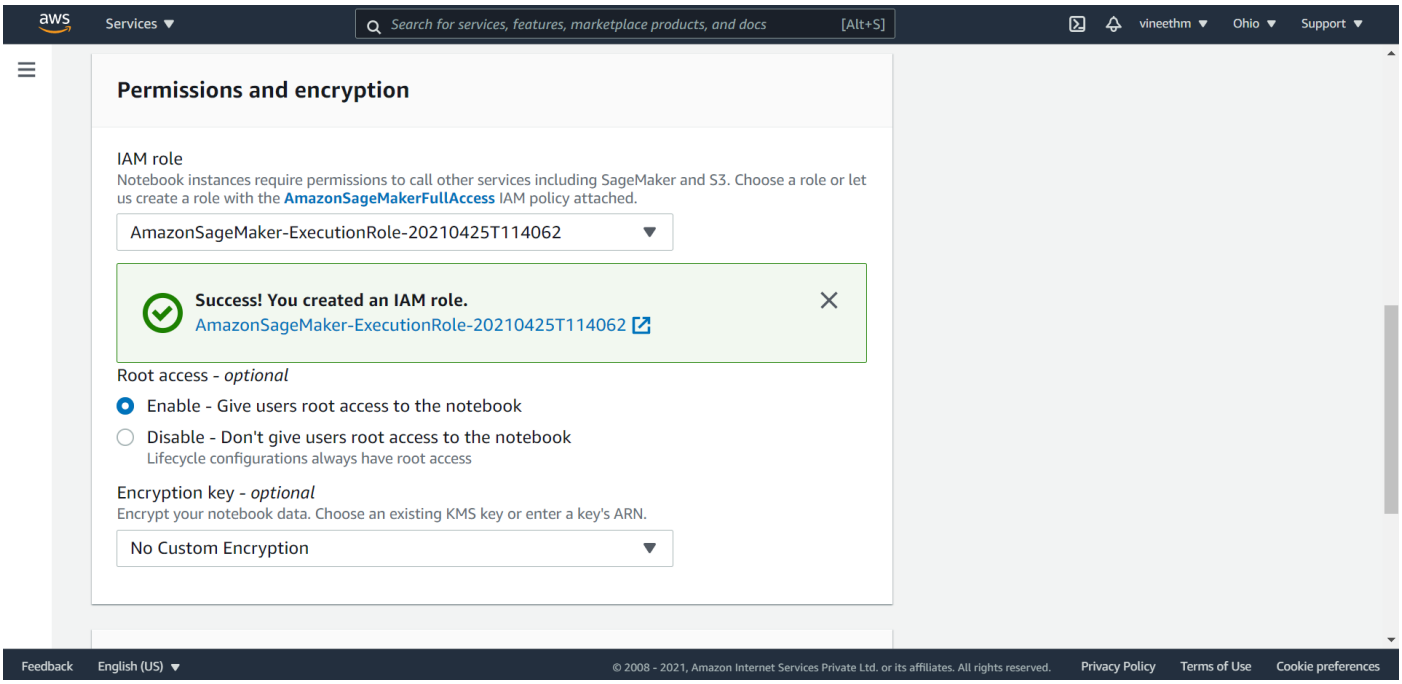
Amazon SageMaker is a fully managed machine learning service. With SageMaker, data scientists and
developers can quickly and easily build and train machine learning models, and then directly deploy
them into a production-ready hosted environment.
 It provides an integrated Jupyter authoring notebook instance for easy access to your data sources for
exploration and analysis, so you don't have to manage servers.
 It also provides common machine learning algorithms that are optimized to run efficiently against
extremely large data in a distributed environment.
 With native support for bring-your-own-algorithms and frameworks, SageMaker offers flexible
distributed training options that adjust to your specific workflows.
 Deploy a model into a secure and scalable environment by launching it with a few clicks from
SageMaker Studio or the SageMaker console.
 Training and hosting are billed by minutes of usage, with no minimum fees and no upfront Commitments

Step-1: Build, train and deploy a Machine Learning model with Amazon SageMaker.

Task-1: Create a notebook instance.

Note: Create an IAM role to enable the SageMaker access S3 buckets.


Create an IAM role
Passing an IAM role gives Amazon SageMaker permission to perform actions in other AWS services on your behalf. Creating a role here will grant permissions described by the AmazonSageMakerFullAccess IAM policy to the role you create.
The IAM role you create will provide access to:
S3 buckets you specify - optional
Any S3 bucket
Allow users that have access to your notebook instance access to any bucket and its contents in your account.
Specific S3 buckets
Example: bucket-name-1, bucke
Comma delimited. ARNs, "*" and "/" are not supported.
None
Any S3 bucket with "sagemaker" in the name
Cancel
Create role


Services
Search for services, features, marketplace products, and docs [Alt+S]
vineethm
Ohio
Support
Permissions and encryption
IAM role
Notebook instances require permissions to call other services including SageMaker and S3. Choose a role or let us create a role with the AmazonSageMakerFullAccess IAM policy attached.
AmazonSageMaker-ExecutionRole-20210425T114062
Success! You created an IAM role.
AmazonSageMaker-ExecutionRole-20210425T114062
Root access - optional
Enable - Give users root access to the notebook
Disable - Don't give users root access to the notebook
Lifecycle configurations always have root access
Encryption key - optional
Encrypt your notebook data. Choose an existing KMS key or enter a key's ARN.
No Custom Encryption
Feedback
English (US)
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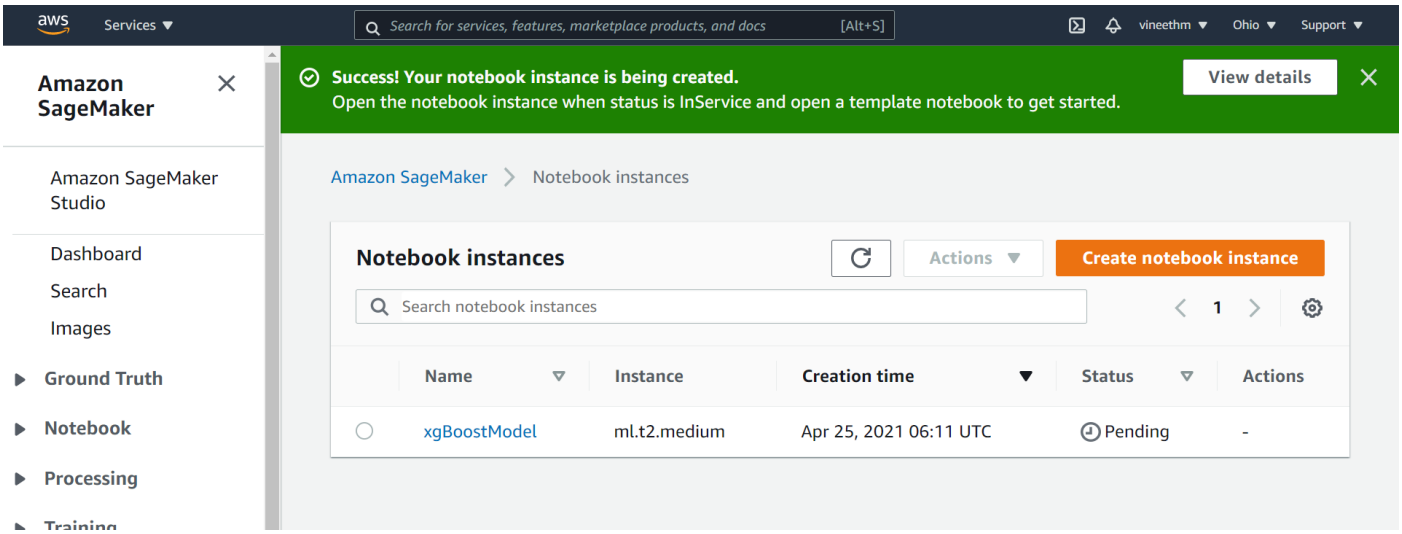
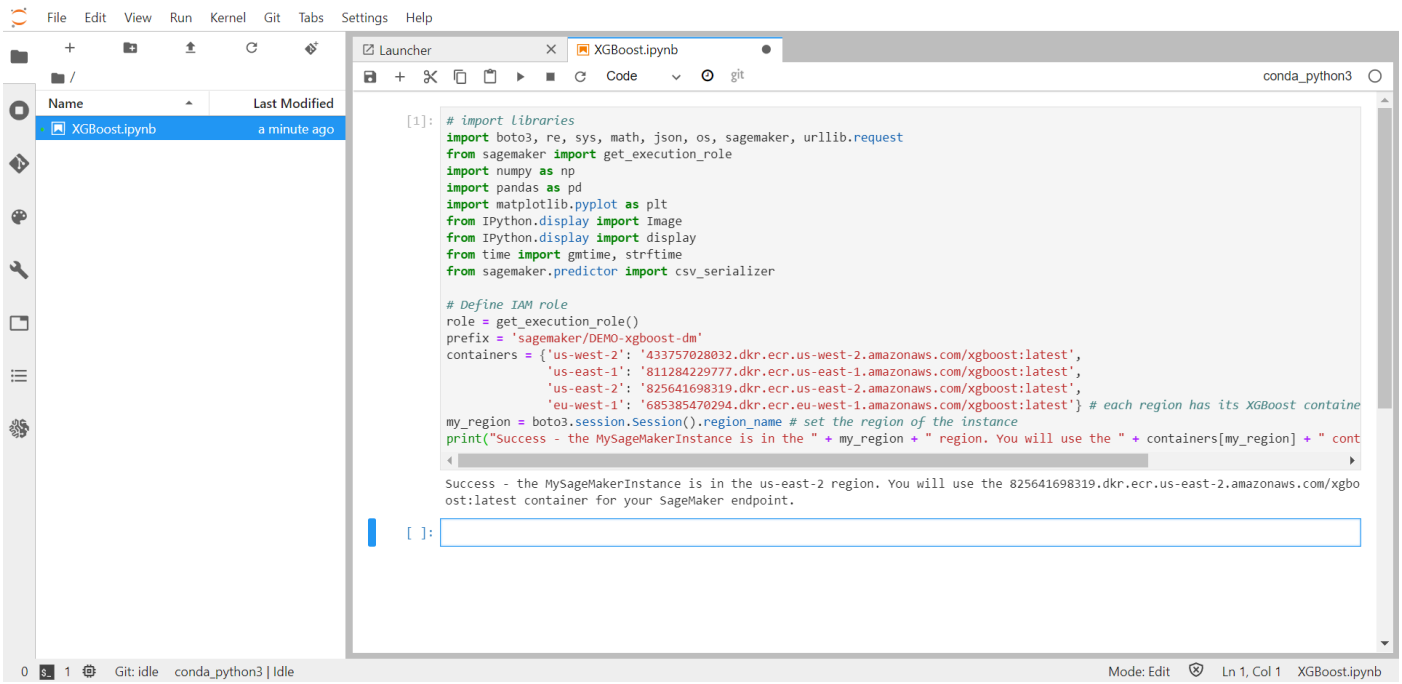
Privacy Policy
Terms of Use
Cookie preferences



Step-2: Preprocess the data in the Jupyter Notebook provisioned by SageMaker.

Task-1: Choose the kernel environment “conda_python3”.




```
[6]: bucket_name = 'sagemaker.xgboost.demo' # <--- CHANGE THIS VARIABLE TO A UNIQUE NAME FOR YOUR BUCKET
s3 = boto3.resource('s3')
try:
    if  my_region == 'us-east-1':
      s3.create_bucket(Bucket=bucket_name)
    else:
      s3.create_bucket(Bucket=bucket_name, CreateBucketConfiguration={ 'LocationConstraint': my_region })
    print('S3 bucket created successfully')
except Exception as e:
    print('S3 error: ',e)

S3 bucket created successfully
```

[7]:
```
try:
  urllib.request.urlretrieve ("https://d1.awsstatic.com/tmt/build-train-deploy-machine-learning-model-sagemaker/bank_clean.27f01
  print('Success: downloaded bank_clean.csv.')
except Exception as e:
  print('Data load error: ',e)

try:
  model_data = pd.read_csv('./bank_clean.csv',index_col=0)
  print('Success: Data loaded into dataframe.')
except Exception as e:
    print('Data load error: ',e)
```

```
Success: downloaded bank_clean.csv.
Success: Data loaded into dataframe.
```

Step-3: Train the Machine Learning Model

[8]:
```
train_data, test_data = np.split(model_data.sample(frac=1, random_state=1729), [int(0.7 * len(model_data))])
print(train_data.shape, test_data.shape)
```

```
(28831, 61) (12357, 61)
```

[10]:
```
pd.concat([train_data['y_yes'], train_data.drop(['y_no', 'y_yes'], axis=1)], axis=1).to_csv('train.csv', index=False, header=Fal
boto3.Session().resource('s3').Bucket(bucket_name).Object(os.path.join(prefix, 'train/train.csv')).upload_file('train.csv')
s3_input_train = sagemaker.inputs.TrainingInput(s3_data='s3://{}/{}/train'.format(bucket_name, prefix), content_type='csv')
```

[11]:
```
sess = sagemaker.Session()
xgb = sagemaker.estimator.Estimator(containers[my_region],role, instance_count=1, instance_type='ml.m4.xlarge',output_path='s3:/
xgb.set_hyperparameters(max_depth=5,eta=0.2,gamma=4,min_child_weight=6,subsample=0.8,silent=0,objective='binary:logistic',num_ro
```

[ ]:

[*]:
```
xgb.fit({'train': s3_input_train})
```

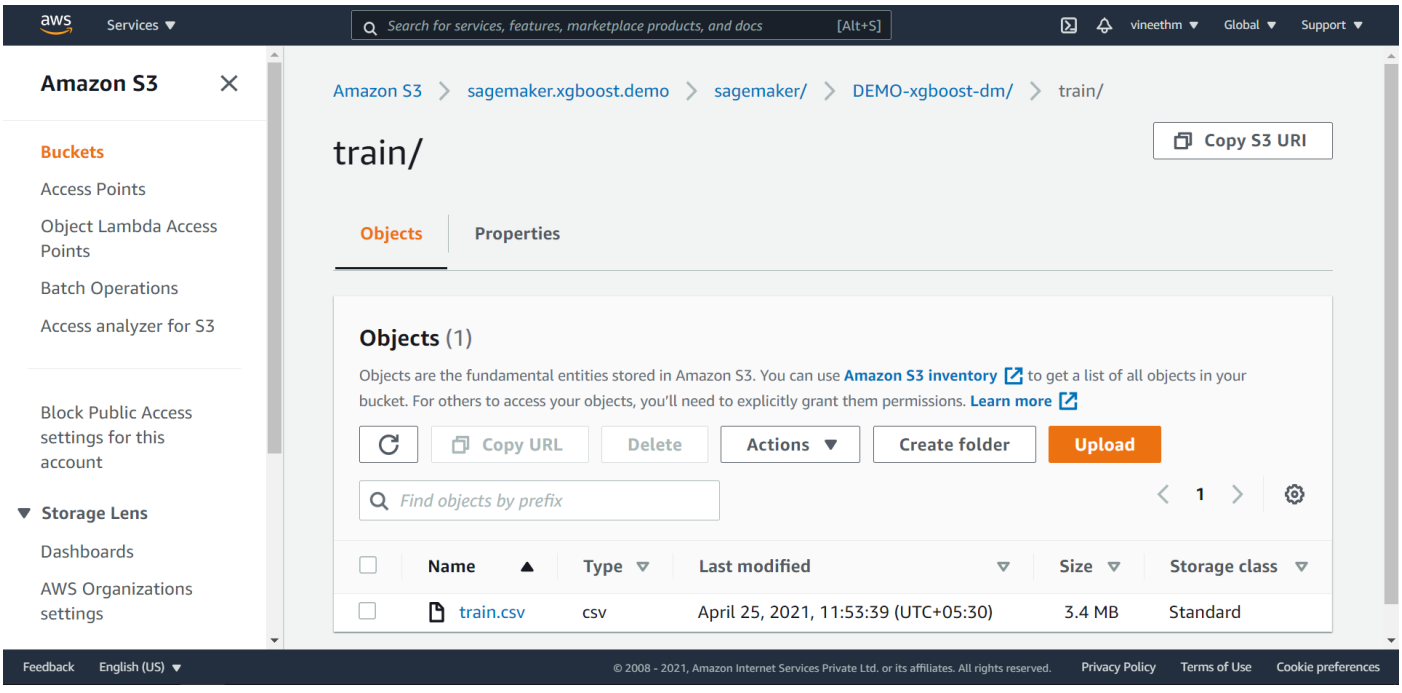
```
2021-04-25 06:24:56 Starting - Starting the training job...
2021-04-25 06:25:20 Starting - Launching requested ML instancesProfilerReport-1619331896: InProgress
.....
2021-04-25 06:26:20 Starting - Preparing the instances for training.......
```

[ ]:

```
[*]: xgb.fit({'train': s3_input_train})

2021-04-25 06:24:56 Starting - Starting the training job...
2021-04-25 06:25:20 Starting - Launching requested ML instancesProfilerReport-1619331896: InProgress
......
2021-04-25 06:26:20 Starting - Preparing the instances for training.........
2021-04-25 06:27:40 Downloading - Downloading input data...
2021-04-25 06:28:24 Training - Training image download completed. Training in progress..Arguments: train
[2021-04-25:06:28:25:INFO] Running standalone xgboost training.
[2021-04-25:06:28:25:INFO] Path /opt/ml/input/data/validation does not exist!
[2021-04-25:06:28:25:INFO] File size need to be processed in the node: 3.38mb. Available memory size in the node: 8418.43mb
[2021-04-25:06:28:25:INFO] Determined delimiter of CSV input is ','
[06:28:25] S3DistributionType set as FullyReplicated
[06:28:25] 28831x59 matrix with 1701029 entries loaded from /opt/ml/input/data/train?format=csv&label_column=0&delimiter=,
[06:28:25] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 30 extra nodes, 14 pruned nodes, max_depth=5
[0]#011train-error:0.100482
[06:28:25] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 30 extra nodes, 14 pruned nodes, max_depth=5
[1]#011train-error:0.099858
[06:28:25] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 28 extra nodes, 22 pruned nodes, max_depth=5
[2]#011train-error:0.099754
[06:28:25] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 22 extra nodes, 14 pruned nodes, max_depth=5
[3]#011train-error:0.099095
[06:28:25] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 28 extra nodes, 12 pruned nodes, max_depth=5
[4]#011train-error:0.098991
[06:28:25] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 32 extra nodes, 14 pruned nodes, max_depth=5
[5]#011train-error:0.099303
[06:28:25] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 24 extra nodes, 18 pruned nodes, max_depth=5
[6]#011train-error:0.099684
[06:28:25] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 30 extra nodes, 22 pruned nodes, max_depth=5
[7]#011train-error:0.09906
[06:28:26] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 28 extra nodes, 20 pruned nodes, max_depth=5
[8]#011train-error:0.098852
[06:28:26] src/tree/updater_prune.cc:74: tree pruning end, 1 roots, 36 extra nodes, 8 pruned nodes, max_depth=5
```

Note: The required files are created in the S3 bucket.



Step-4: Deploy Model

```
2021-04-25 06:28:41 Uploading - Uploading generated training model
2021-04-25 06:28:41 Completed - Training job completed
Training seconds: 61
Billable seconds: 61
```

[13]:
```
xgb_predictor = xgb.deploy(initial_instance_count=1,instance_type='ml.m4.xlarge')
```

```
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```

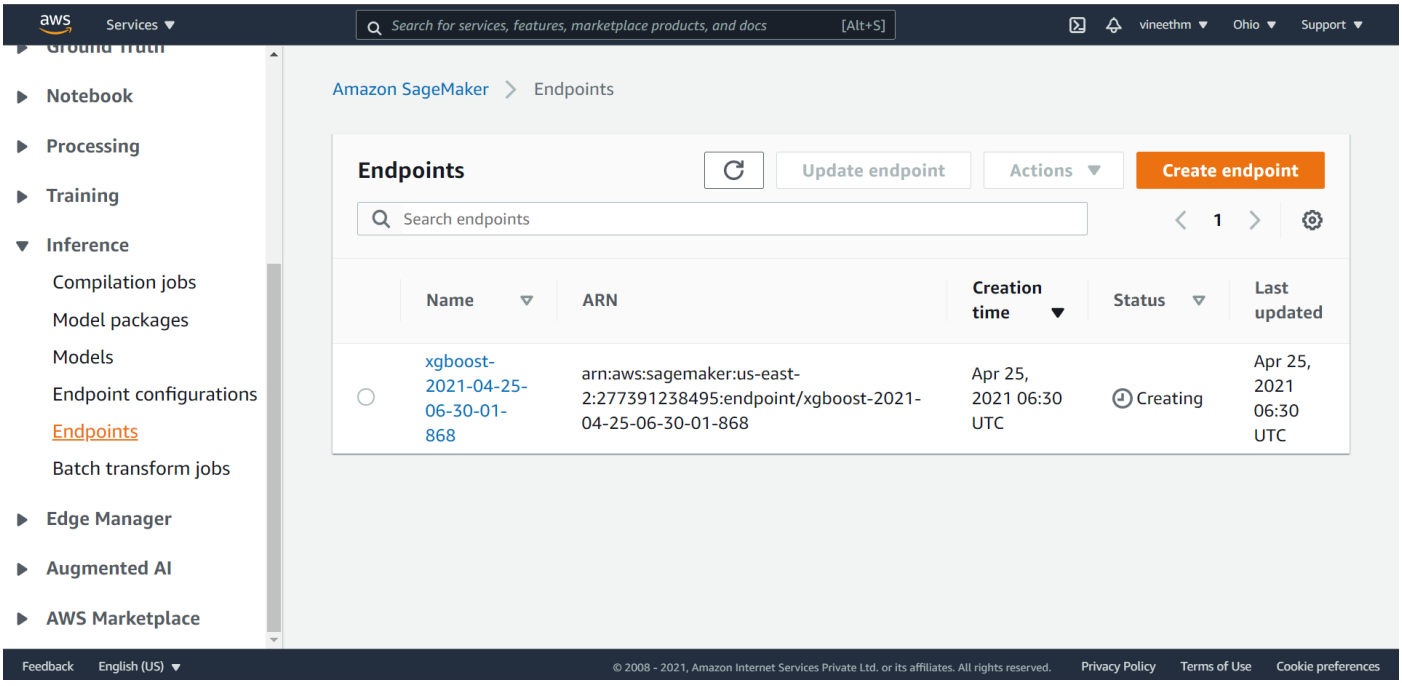
[14]:
```
from sagemaker.serializers import CSVSerializer

test_data_array = test_data.drop(['y_no', 'y_yes'], axis=1).values #load the data into an array
xgb_predictor.serializer = CSVSerializer() # set the serializer type
predictions = xgb_predictor.predict(test_data_array).decode('utf-8') # predict!
predictions_array = np.fromstring(predictions[1:], sep=',') # and turn the prediction into an array
print(predictions_array.shape)
```

```
(12357,)
```

[ ]:

We can notice the endpoint being created.



Step-5: Evaluate Model performance.

```
[15]: cm = pd.crosstab(index=test_data['y_yes'], columns=np.round(predictions_array), rownames=['Observed'], colnames=['Predicted'])
tn = cm.iloc[0,0]; fn = cm.iloc[1,0]; tp = cm.iloc[1,1]; fp = cm.iloc[0,1]; p = (tp+tn)/(tp+tn+fp+fn)*100
print("\n{0:<20}{1:<4.1f}%\n".format("Overall Classification Rate: ", p))
print("{0:<15}{1:<15}{2:>8}".format("Predicted", "No Purchase", "Purchase"))
print("Observed")
print("{0:<15}{1:<2.0f}% ({2:<}){3:>6.0f}% ({4:<})".format("No Purchase", tn/(tn+fn)*100,tn, fp/(tp+fp)*100, fp))
print("{0:<16}{1:<1.0f}% ({2:<}){3:>7.0f}% ({4:<}) \n".format("Purchase", fn/(tn+fn)*100,fn, tp/(tp+fp)*100, tp))
```

```
Overall Classification Rate: 89.5%

Predicted      No Purchase    Purchase
Observed
No Purchase    90% (10769)    37% (167)
Purchase        10% (1133)     63% (288)
```

```
[ ]:
```

Step-6: Clean up the resources.