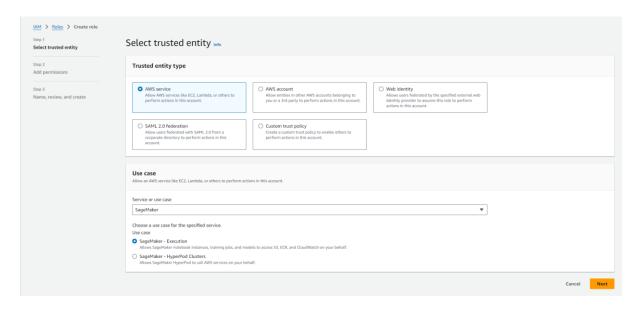
CLOUD COMPUTING PRACTICAL 8:AMAZON SAGEMAKER

Name: Aditi Kothawade

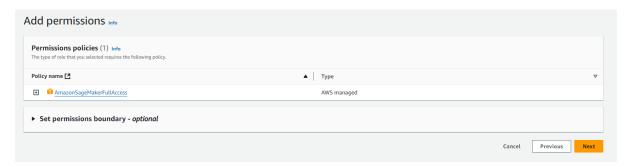
Roll No: A074

MSc SDS Batch 2

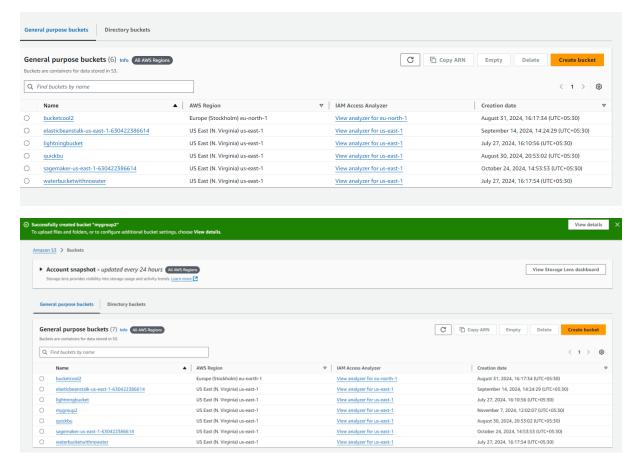
1)Creating IAM ROLE and assigning sagemaker permission



IAM Role is created.



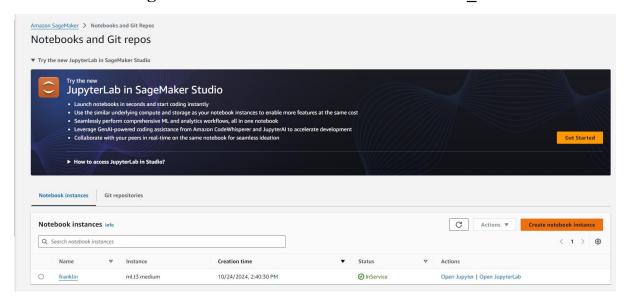
2) creating s3 bucket named mygroup2



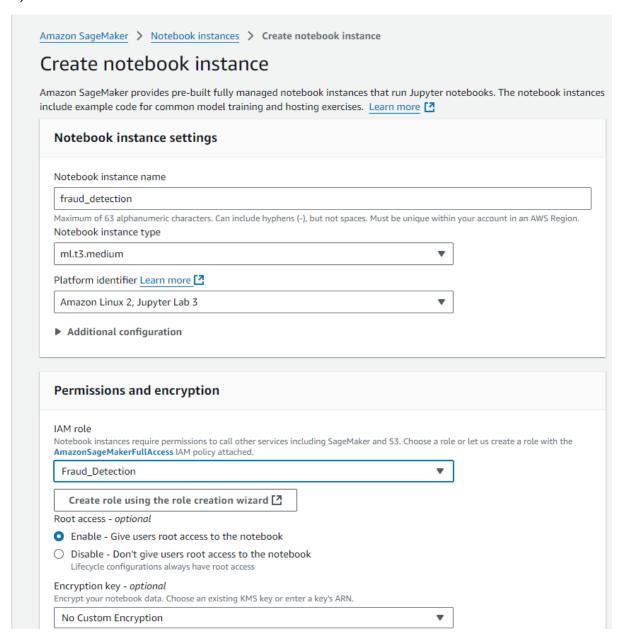
3) open Amazon SageMaker console

Select Notebook instances and click create notebook instances

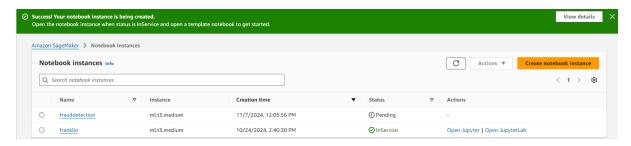
Here we will assign the IAM role created earlier i.e fraud detection



4) CREATE A JUPYTER NOTEBOOK

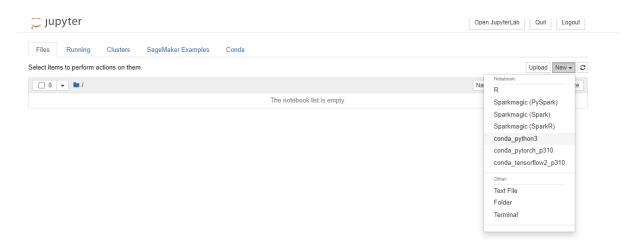


Notebook is created



1. Open Jupyter or JupyterLab according to the interface needed.

- 2. Go to File menu->Choose New-> Notebook.
- 3. Select Kernel as 'conda_python3'



Deploying the model (Here it is stored in s3 bucket that we had created)

```
In [1]: import shap
X, y = shap.datasets.adult()
              X_display, y_display = shap.datasets.adult(display=True)
              feature_names = list(X.columns)
              feature names
              Matplotlib is building the font cache; this may take a moment.
   Out[1]: ['Age',
                'Workelass'
               'Education-Num'
               'Marital Status',
                'Occupation',
                'Relationship'
                'Race',
               'Sex',
'Capital Gain',
                'Capital Loss'
                'Hours per week',
               'Country']
In [7]: import sagemaker, boto3, os
          bucket = sagemaker.Session().default_bucket()
prefix = "demo-sagemaker-xgboost-adult-income-prediction"
          boto3.Session().resource('s3').Bucket(bucket).Object(
   os.path.join(prefix, 'data/train.csv')).upload_file('train.csv')
boto3.Session().resource('s3').Bucket(bucket).Object(
   os.path.join(prefix, 'data/validation.csv')).upload_file('validation.csv')
          sagemaker.config INFO - Not applying SDK defaults from location: /etc/xdg/sagemaker/config.yaml
          sagemaker.config INFO - Not applying SDK defaults from location: /home/ec2-user/.config/sagemaker/config.yaml
          region = sagemaker.Session().boto_region_name
          print("AWS Region: {}".format(region))
          role = sagemaker.get_execution_role()
          AWS Region: us-east-1
          RoleArn: arn:aws:iam::975050009706:role/lucifer007
```

```
! aws s3 cp {rule_output_path} ./ --recursive

from IPython.display import FileLink, FileLinks
display("Click link below to view the XGBoost Training report", FileLink("CreateXgboostReport/xgboost_report.html"))

download: s3://sagemaker-us-east-1-975050009706/demo-sagemaker-xgboost-adult-income-prediction/xgboost_model/sagemaker-xgboost-2024-10-24-09-29-24-130/rule-output/CreateXgboostReport/xgboost-reports/EvaluationMetrics.json to CreateXgboostReport/xgboost-reports/EvaluationMetrics.json to CreateXgboostReport/xgboost-adult-income-prediction/xgboost_model/sagemaker-xgboost-2024-10-24-09-29-24-130/rule-output/CreateXgboostReport/xgboost-reports/FeatureImportance.json to CreateXgboostReport/xgboost-reports/FeatureImportance.json to CreateXgboostReport/xgboost-reports/FeatureImportance.json to CreateXgboostReport/xgboost-reports/FeatureImportance.json to CreateXgboostReport/xgboost-adult-income-prediction/xgboost_model/sagemaker-xgboost-2024-10-24-09-29-24-130/rule-output/ProfilerReport/profiler-output/profiler-report.ipynb to ProfilerReport/profiler-output/profiler-report.ipynb download: s3://sagemaker-us-east-1-975050009706/demo-sagemaker-xgboost-adult-income-prediction/xgboost_model/sagemaker-xgboost-2024-10-24-09-29-24-130/rule-output/CreateXgboostReport/xgboost-reports/ConfusionMatrix.json to CreateXgboostReport/xgboost-reports/ConfusionMatrix.json to CreateXgboostReport/xgboost-reports/ConfusionMatrix.json
```

```
from sagemaker.debugger import Rule, ProfilerRule, rule_configs
from sagemaker.session import TrainingInput
s3_output_location='s3://{}/{}/.format(bucket, prefix, 'xgboost_model')
container=sagemaker.image_uris.retrieve("xgboost", region, "1.2-1")
print(container)
xgb_model=sagemaker.estimator.Estimator(
    image_uri=container,
    role=role,
    instance count=1,
    instance_type='ml.m4.xlarge',
    volume size=5,
    output_path=s3_output_location,
    sagemaker_session=sagemaker.Session(),
    rules=[
        Rule.sagemaker(rule_configs.create_xgboost_report()),
        ProfilerRule.sagemaker(rule_configs.ProfilerReport())
)
```

```
In [18]: xgb_predictor.endpoint_name
Out[18]: 'sagemaker-xgboost-2024-10-24-09-34-02-816'
predictions = for array in split_array:
    predictions = ','.join([predictions, xgb_predictor.predict(array).decode('utf-8')])
return np.fromstring(predictions[1:], sep=',')
In [20]: import matplotlib.pyplot as plt
         predictions=predict(test.to_numpy()[:,1:])
         plt.hist(predictions)
         plt.show()
          3500
          3000
          2500
          2000
          1500
           1000
            500
                                         0.4
                  0.0
                             0.2
                                                    0.6
                                                                0.8
                                                                            1.0
```

```
In [21]: import sklearn
           cutoff=0.5
           print(sklearn.metrics.confusion_matrix(test.iloc[:, 0], np.where(predictions > cutoff, 1, 0))) print(sklearn.metrics.classification_report(test.iloc[:, 0], np.where(predictions > cutoff, 1, 0)))
           [[4670 356]
             [ 480 1007]]
                              precision
                                              recall f1-score support
                          0
                                    0.91
                                                 0.93
                                                              0.92
                                                                          5026
                                    0.74
                                                0.68
                                                              0.71
                                                                          1487
                 accuracy
                                                              0.87
                                                                          6513
               macro avg
                                    0.82
                                                 0.80
                                                              0.81
                                                                          6513
           weighted avg
                                    0.87
                                                0.87
                                                              0.87
                                                                          6513
```

