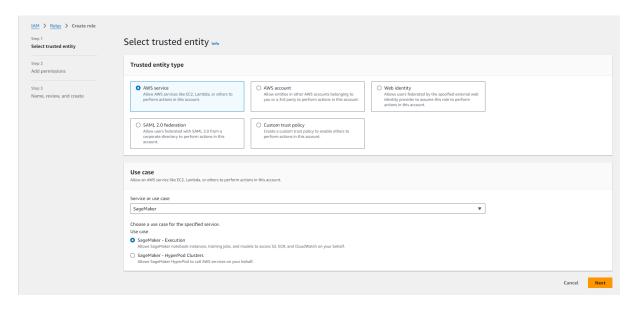
# CLOUD COMPUTING PRACTICAL 8:AMAZON SAGEMAKER

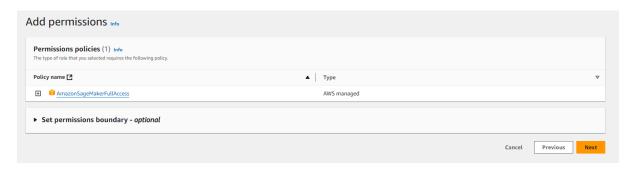
# **Shriomi Gite**

# **A015**

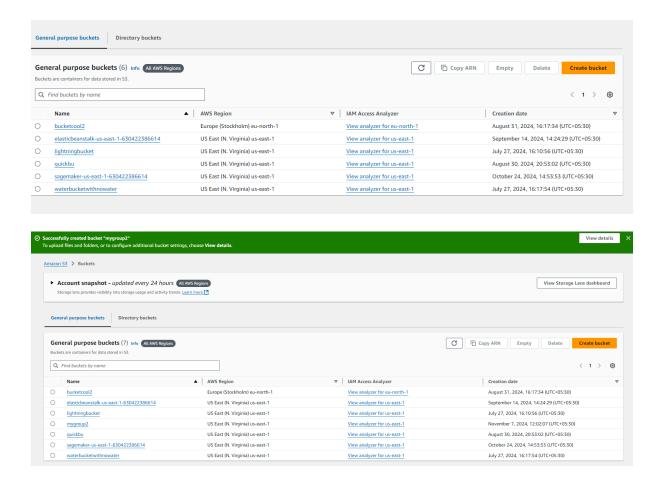
# 1)Creating IAM ROLE and assigning sagemaker permission



### IAM Role is created.



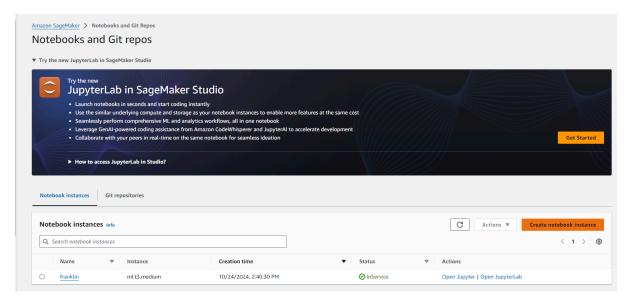
2) creating s3 bucket named mygroup2



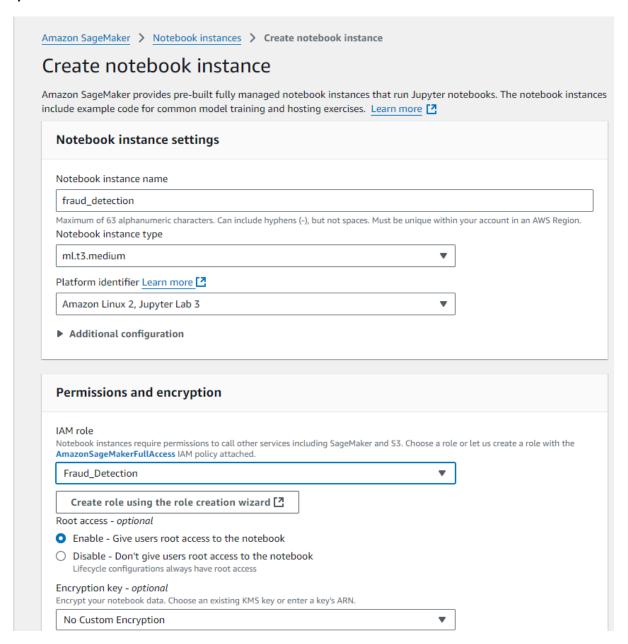
#### 3) open Amazon SageMaker console

#### Select Notebook instances and click create notebook instances

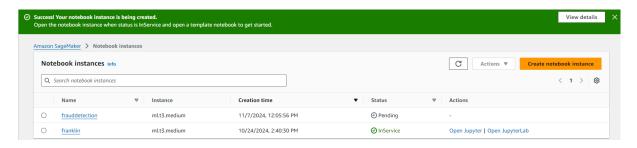
Here we will assign the IAM role created earlier i.e fraud\_detection



#### 4) CREATE A JUPYTER NOTEBOOK

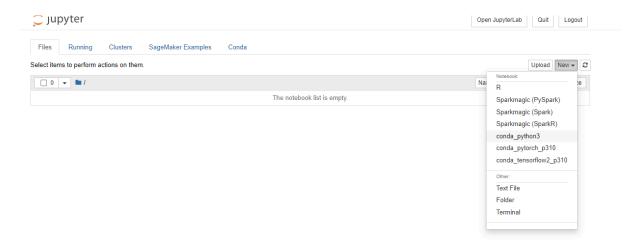


## Notebook is created



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)
                Matplotlib is building the font cache; this may take a moment.
   Out[1]: ['Age',
'Workclass',
                  '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
In [8]: import sagemaker
            region = sagemaker.Session().boto_region_name
            print("AWS Region: {}".format(region))
            role = sagemaker.get_execution_role()
            print("RoleArn: {}".format(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

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/FeatureImportance.json to CreateXgboostReport/xgboost-reports/FeatureImportance.json

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/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

```
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'
In [19]: import numpy as np
def predict(data, rows=1000):
    split_array = np.array_split(data, int(data.shape[0] / float(rows) + 1))
    predictions = ''
    for a predy in split array:
                   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
                                                           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
                           1
                                     0.74
                                                  0.68
                                                                0.71
                                                                             1487
                 accuracy
                                                                0.87
                                                                             6513
                                     0.82
                                                  0.80
                macro avg
                                                                0.81
                                                                             6513
            weighted avg
                                     0.87
                                                  0.87
                                                                0.87
                                                                             6513
```

