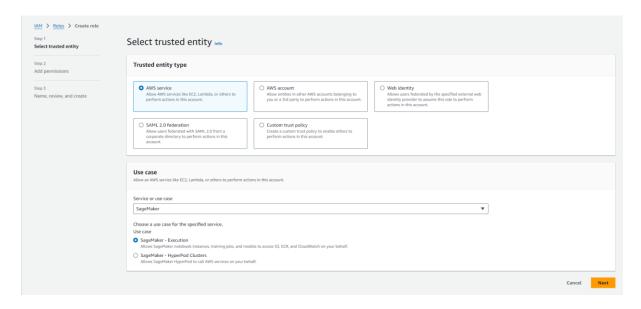
# CLOUD COMPUTING PRACTICAL 8:AMAZON SAGEMAKER

Name: Akhilesh Nambiar

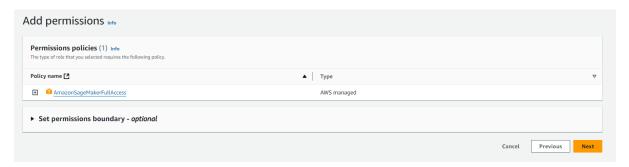
Roll No: A043

MSc SDS Batch 2

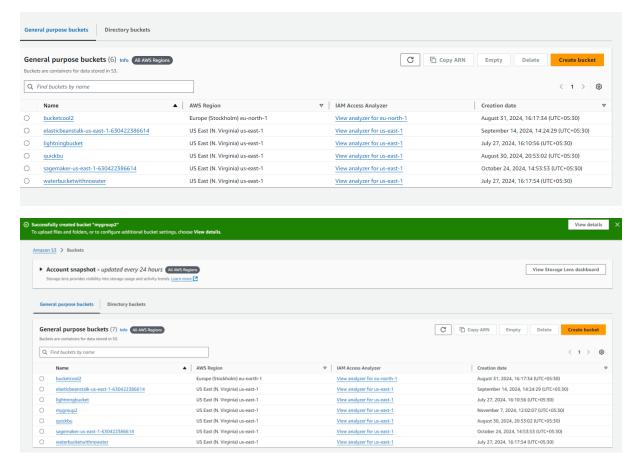
## 1)Creating IAM ROLE and assigning sagemaker permission



#### IAM Role is created.



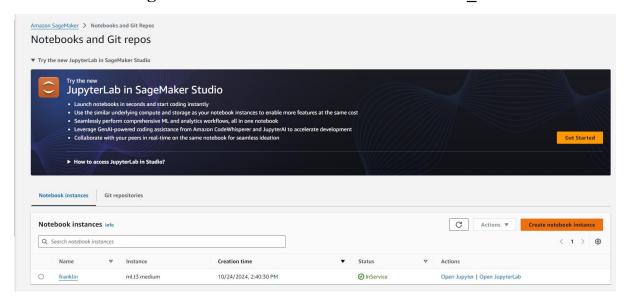
2) creating s3 bucket named mygroup2



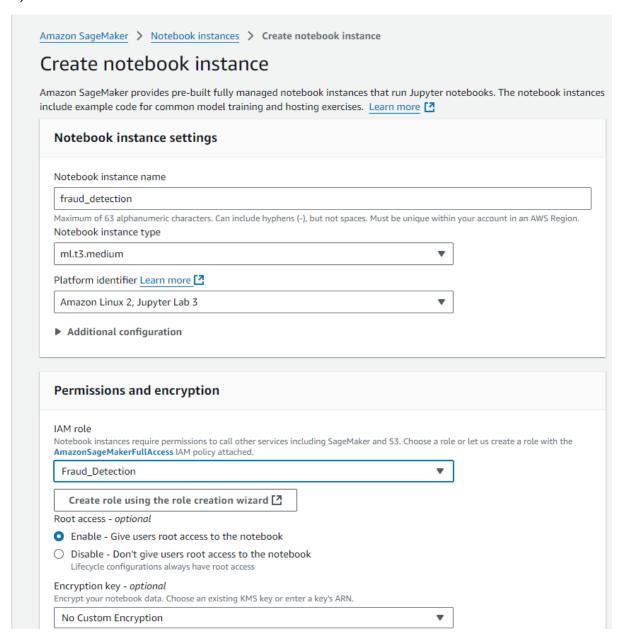
# 3) open Amazon SageMaker console

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

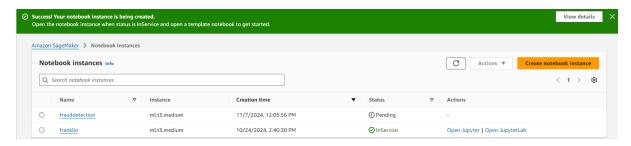
### Here we will assign the IAM role created earlier i.e fraud detection



# 4) CREATE A JUPYTER NOTEBOOK

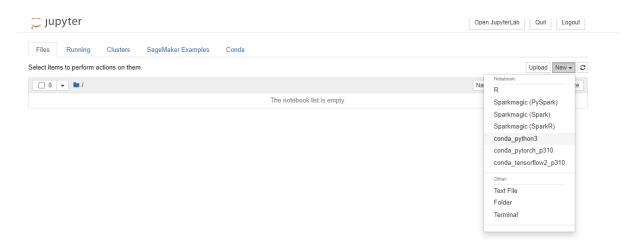


#### Notebook is created



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
```

