

SVKM'S NMIMS Nilkamal School of Mathematics, Applied Statistics & Analytics

Master of Science (Statistics & Data Science)

PRACTICAL 8:AMAZON SAGEMAKER

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1)Creating IAM ROLE and assigning sagemaker permission

The screenshot shows the 'Create role' page in the AWS IAM console. The left sidebar indicates the current step is 'Step 1: Select trusted entity'. The main content area is titled 'Select trusted entity' and includes a 'Trusted entity type' section with five options: 'AWS service' (selected), 'AWS account', 'Web identity', 'SAML 2.0 federation', and 'Custom trust policy'. Below this is a 'Use case' section with a dropdown menu set to 'SageMaker' and two use case options: 'SageMaker - Execution' (selected) and 'SageMaker - HyperPod Clusters'. At the bottom right are 'Cancel' and 'Next' buttons.

IAM Role is created.

The screenshot shows the 'Add permissions' page in the AWS IAM console. The left sidebar indicates the current step is 'Step 2: Add permissions'. The main content area is titled 'Add permissions' and includes a 'Permissions policies (1)' section with a table listing the selected policy: 'AmazonSageMakerFullAccess' of type 'AWS managed'. Below the table is a section for 'Set permissions boundary - optional'. At the bottom right are 'Cancel', 'Previous', and 'Next' buttons.

2) creating s3 bucket named mygroup2

General purpose buckets | Directory buckets

General purpose buckets (6) [Info](#) [All AWS Regions](#)

Buckets are containers for data stored in S3.

Find buckets by name

Name	AWS Region	IAM Access Analyzer	Creation date
bucketcool2	Europe (Stockholm) eu-north-1	View analyzer for eu-north-1	August 31, 2024, 16:17:34 (UTC+05:30)
elasticbeanstalk-us-east-1-630422386614	US East (N. Virginia) us-east-1	View analyzer for us-east-1	September 14, 2024, 14:24:29 (UTC+05:30)
lightningbucket	US East (N. Virginia) us-east-1	View analyzer for us-east-1	July 27, 2024, 16:10:56 (UTC+05:30)
quickbu	US East (N. Virginia) us-east-1	View analyzer for us-east-1	August 30, 2024, 20:53:02 (UTC+05:30)
sagemaker-us-east-1-630422386614	US East (N. Virginia) us-east-1	View analyzer for us-east-1	October 24, 2024, 14:53:53 (UTC+05:30)
waterbucketwithnowater	US East (N. Virginia) us-east-1	View analyzer for us-east-1	July 27, 2024, 16:17:54 (UTC+05:30)

Successfully created bucket "mygroup2" [View details](#)

To upload files and folders, or to configure additional bucket settings, choose [View details](#).

Amazon S3 > Buckets

Account snapshot - updated every 24 hours [All AWS Regions](#) [View Storage Lens dashboard](#)

Storage lens provides visibility into storage usage and activity trends. [Learn more](#)

General purpose buckets | Directory buckets

General purpose buckets (7) [Info](#) [All AWS Regions](#)

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lightningbucket	US East (N. Virginia) us-east-1	View analyzer for us-east-1	July 27, 2024, 16:10:56 (UTC+05:30)
mygroup2	US East (N. Virginia) us-east-1	View analyzer for us-east-1	November 7, 2024, 12:02:07 (UTC+05:30)
quickbu	US East (N. Virginia) us-east-1	View analyzer for us-east-1	August 30, 2024, 20:53:02 (UTC+05:30)
sagemaker-us-east-1-630422386614	US East (N. Virginia) us-east-1	View analyzer for us-east-1	October 24, 2024, 14:53:53 (UTC+05:30)
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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

Amazon SageMaker > Notebooks and Git Repos

Notebooks and Git repos

Try the new JupyterLab in SageMaker Studio

Try the new JupyterLab in SageMaker Studio

- Launch notebooks in seconds and start coding instantly
- Use the similar underlying compute and storage as your notebook instances to enable more features at the same cost
- Seamlessly perform comprehensive ML and analytics workflows, all in one notebook
- Leverage GenAI-powered coding assistance from Amazon CodeWhisperer and JupyterAI to accelerate development
- Collaborate with your peers in real-time on the same notebook for seamless ideation

[Get Started](#)

How to access JupyterLab in Studio?

Notebook instances | Git repositories

Notebook instances [Info](#)

Search notebook instances

Name	Instance	Creation time	Status	Actions
franklin	ml.t3.medium	10/24/2024, 2:40:30 PM	InService	Open Jupyter Open JupyterLab

4) CREATE A JUPYTER NOTEBOOK

Create notebook instance

Amazon SageMaker provides pre-built fully managed notebook instances that run Jupyter notebooks. The notebook instances include example code for common model training and hosting exercises. [Learn more](#)

Notebook instance settings

Notebook instance name

fraud_detection

Maximum of 63 alphanumeric characters. Can include hyphens (-), but not spaces. Must be unique within your account in an AWS Region.

Notebook instance type

ml.t3.medium

Platform identifier [Learn more](#)

Amazon Linux 2, Jupyter Lab 3

► Additional configuration

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.

Fraud_Detection

Create role using the role creation wizard

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

Notebook is created

Success! Your notebook instance is being created.

Open the notebook instance when status is InService and open a template notebook to get started.

[View details](#)

Amazon SageMaker > Notebook instances

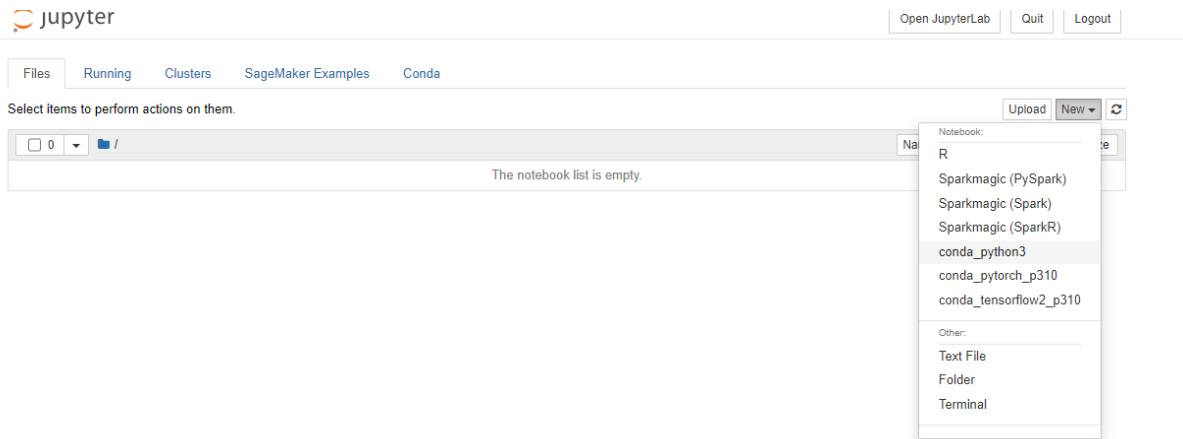
Notebook instances Info

Search notebook instances

Name	Instance	Creation time	Status	Actions
frauddetection	ml.t3.medium	11/7/2024, 12:05:56 PM	Pending	-
franklin	ml.t3.medium	10/24/2024, 2:40:30 PM	InService	Open Jupyter Open JupyterLab

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',
'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())
    ]
)
```

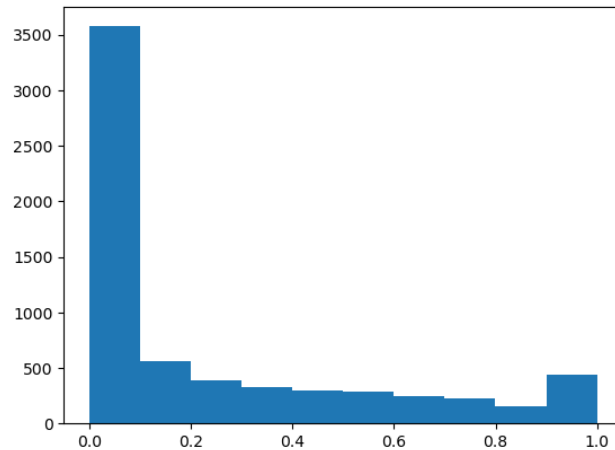
```
In [18]: xgb_predictor.endpoint_name
```

```
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 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:,1:])
plt.hist(predictions)
plt.show()
```



```
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          0.82
 macro avg          0.80
weighted avg          0.87
```

Upload [Info](#)

Add the files and folders you want to upload to S3. To upload a file larger than 160GB, use the AWS CLI, AWS SDK or Amazon S3 REST API. [Learn more](#)

Drag and drop files and folders you want to upload here, or choose **Add files** or **Add folder**.

Files and folders (1 Total, 98.2 MB)

Remove

Add files

Add folder

All files and folders in this table will be uploaded.

Q creditcard X

1 match < 1 >

<input type="checkbox"/>	Name	Folder
<input type="checkbox"/>	creditcard.csv	-

Destination [Info](#)

Destination

[s3://mygroup2](#)

► Destination details

Bucket settings that impact new objects stored in the specified destination.

- Permissions

Grant public access and access to other AWS accounts.
- Properties

Specify storage class, encryption settings, tags, and more.

Cancel

Upload