

CSCE 5300: Intro. to Big Data and Data Science Project Fraud Detection in Financial Transactions

Team members

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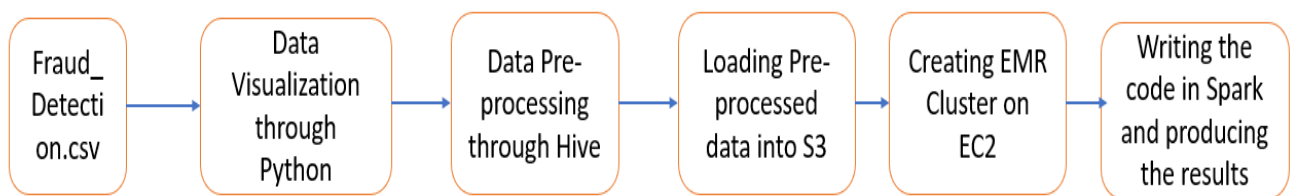
Sathwika Karingu

Krishna Anuhya Regalla

Problem Statement:

Reducing the financial impact of credit card fraud, identity theft, and payment fraud requires constructing a robust fraud detection system. Big data analysis requires the use of sophisticated tools like Hadoop, Spark, and Hive because to the significant increase in digital transactions and the resulting expansion in data. A proficient system must incorporate data processing, machine learning, and real-time analytics to effectively prevent fraud.

Workflow Diagram:



Tools used in throughout the project :

- **AWS**
- **HIVE**

Implementation Status Report :

Description :

We are a team of 4 people working on the fraud transactions detection model. We have implemented Random Forest classifier to train our model in pyspark and Python.

Responsibility:

Sathwika – (Team Lead) Helped in building the EMR cluster and working throughout the Notebook.

Anuhya – Worked on the Data Visualization

Vivek – Data Preprocessing through Hive.

Hari - Writing the code for training the model and Hyper tuning parameters.

Contributions:

Sathwika took 25 percent of the project contribution in making the report and the presentation for the project. She is responsible with AWS and Cloudera setup.

Anuhya took 25 percent of the project contribution in making the report and the presentation for the project and also performed Data Visualization using python until the data Preprocessing steps

Vivek- took 25 percent of the project contribution in making the report and the presentation for the project and have written code in SQL.

Hari - took 25 percent of the project contribution in making the report and the presentation for the project and have written code in Python.

Issues/ Concerns:

We could have performed more on Hive. We have rather focused on Machine Learning and testing the model instead of working more on preprocessing part. That's our major concern in this project.

Walking through the Data set :

Step – It's a integer Data Type which refers to transactions numbering

Type – It's a String Data Type which contains the payment mode

Amount – Integer or Float value with amount

nameOrig - This column likely represents the originating account or entity's name or identifier. It identifies the source of the transaction.

oldbalanceOrg – Integer value with amount before transaction

newbalanceOrig: Following the transaction, the balance in the originating account is shown in this column. It displays the account balance following a payment or transfer of funds.

nameDest: This field most likely contains the name or identification of the destination account or entity. It indicates who the transaction's recipient or destination is.

oldbalanceDest: The balance in the destination account prior to the transaction is shown in this column. It shows the amount in the recipient's account prior to the funds being sent.

newbalanceDest: Following the transaction, the balance in the destination account is shown in this column. It displays the recipient's account balance following the money transfer.

isFraud- An single integer to state whether the transaction is Fraud or not

isFlaggedFraud – This is also similar to the above variable, It's a potential

additional fraud detection measure.

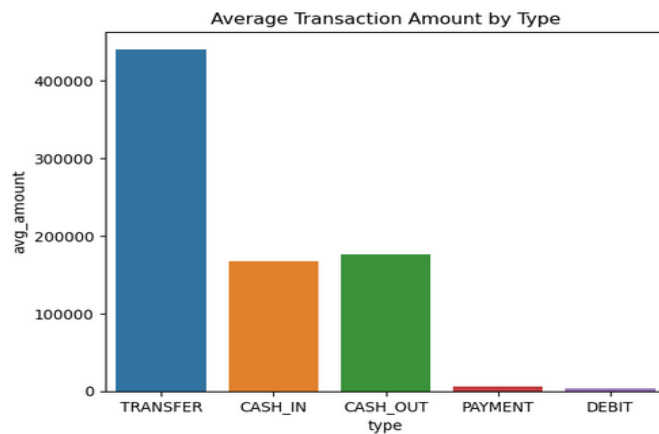
Data Visualization in Python:

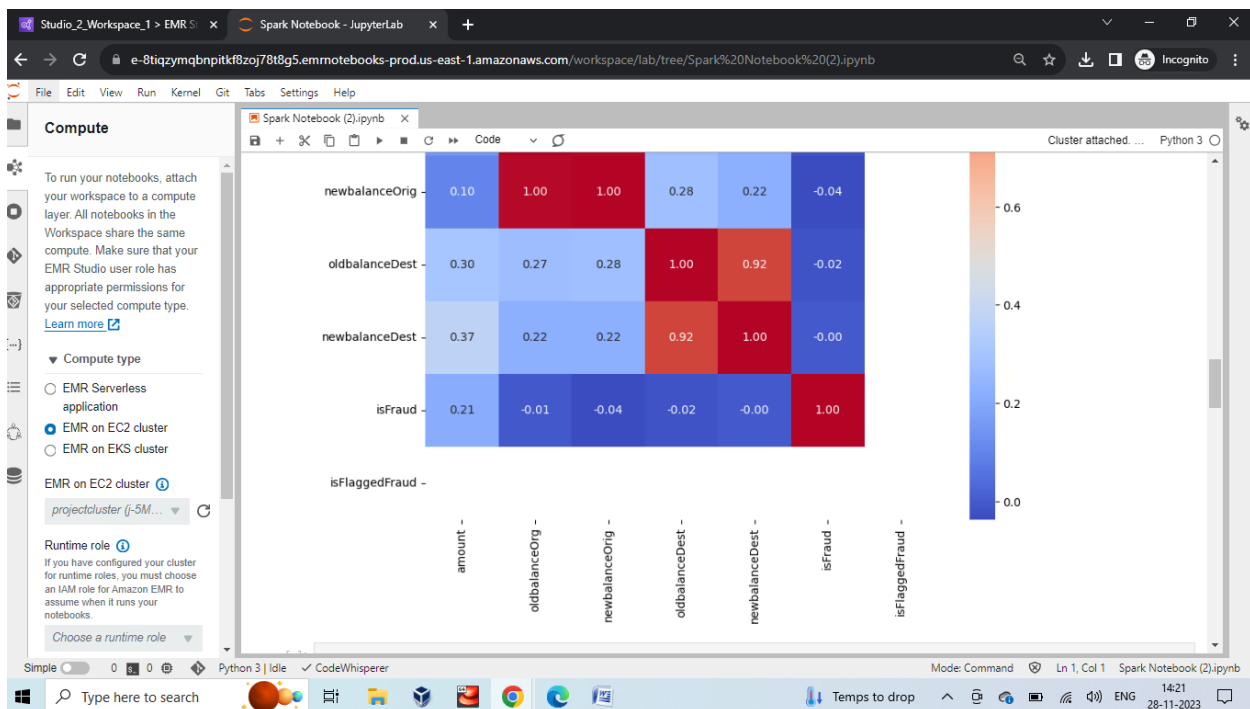
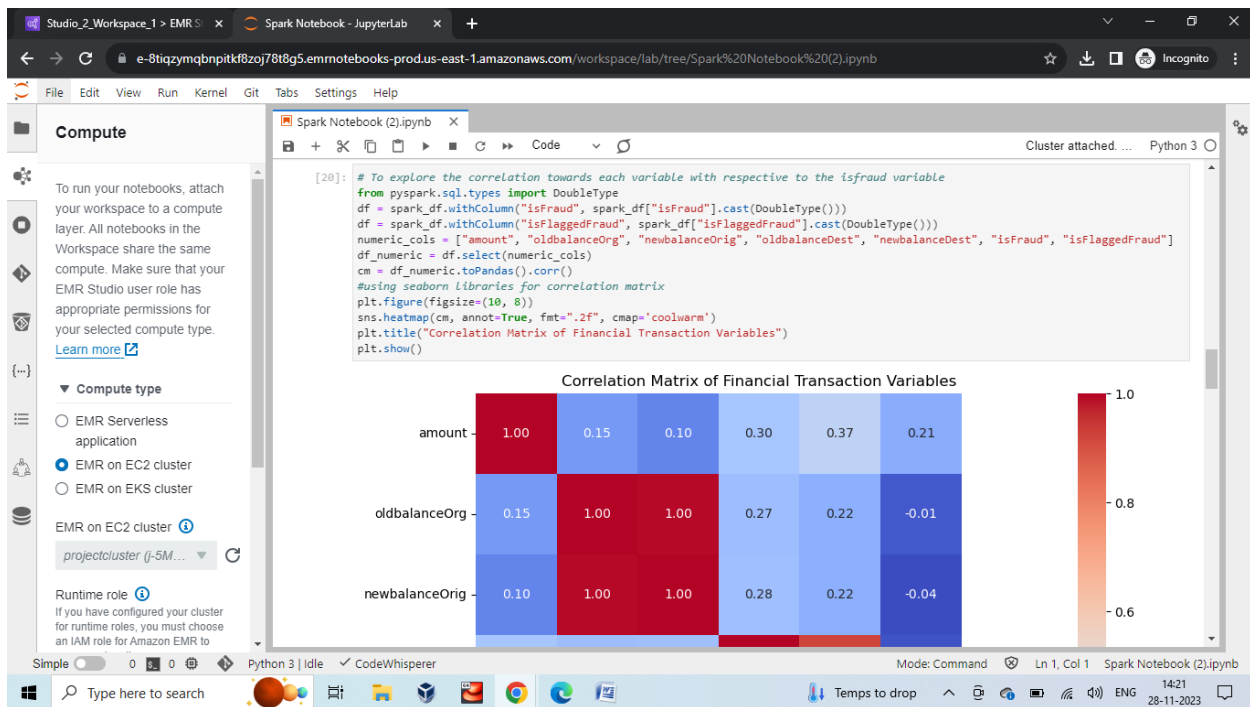
```
In [18]: #Data pre-processing
selected_cols = ["amount", "oldbalanceOrig", "newbalanceOrig", "oldbalanceDest", "newbalanceDest", "isFraud"]
data = spark_df.select(selected_cols)
data = data.dropna()
data = data.withColumn("label", lit(1.0).cast("double"))
```

```
In [19]: #Data Visualization to understand the relationship in between the variables.
from pyspark.sql import functions as F

type_vs_amount = spark_df.groupBy("type").agg(
    F.avg("amount").alias("avg_amount"),
    F.sum("amount").alias("total_amount")
).toPandas()
#after converting to pandas dataframe i have used bar chart to compare the values
import matplotlib.pyplot as plt
import seaborn as sns

sns.barplot(x='type', y='avg_amount', data=type_vs_amount)
plt.title('Average Transaction Amount by Type')
plt.show()
#Looks Like transctions with "transfers" are more in number.
```





Data Pre-processing in Hive :

Step -01 :

Setup Hive in the Cloudera Platform and create a empty table with the schema as it is present in the csv file.

Code to load the data from CSV to Hive Table:

```
CREATE TABLE FraudTransactions (step INT,type STRING,amount DOUBLE,  
    nameOrig STRING,  
    oldbalanceOrg DOUBLE,  
    newbalanceOrig DOUBLE,  
    nameDest STRING,  
    oldbalanceDest DOUBLE,  
    newbalanceDest DOUBLE,  
    isFraud INT,  
    isFlaggedFraud INT  
)  
  
ROW FORMAT DELIMITED  
  
FIELDS TERMINATED BY ','  
  
STORED AS TEXTFILE;
```

Output:

cloudera-quickstart-vm-5.13.0-0-virtualbox [Running] - Oracle VM VirtualBox

File Machine View Input Devices Help

Applications Places System Mon Nov 27, 12:18 PM cloudera

```
cloudera@quickstart:~  
File Edit View Search Terminal Help  
> CREATE TABLE FraudTransactions (step INT,type STRING,amount DOUBLE,  
> nameOrig STRING,  
> oldbalanceOrg DOUBLE,  
> newbalanceOrig DOUBLE,  
> nameDest STRING,  
> oldbalanceDest DOUBLE,  
> newbalanceDest DOUBLE,  
> isFraud INT,  
> isFlaggedFraud INT  
> )  
> ROW FORMAT DELIMITED  
> FIELDS TERMINATED BY ','  
> STORED AS TEXTFILE;  
OK  
Time taken: 0.105 seconds  
hive> load data local inpath '/home/cloudera/Downloads/Fraud_detection_system  
.csv' into table FraudTransactions;  
Loading data to table default.fraudtransactions  
Table default.fraudtransactions stats: [numFiles=1, totalSize=391964]  
OK  
Time taken: 0.518 seconds  
hive> select * from FraudTransactions;
```

[Termin... [Trash] cloudera... [Search ... [cloudera]

Type here to search 52°F 14:18 27-11-2023

cloudera-quickstart-vm-5.13.0-0-virtualbox [Running] - Oracle VM VirtualBox

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```
cloudera@quickstart:~  
File Edit View Search Terminal Help  
6 PAYMENT 8545.54 C1242315432 20189.12 11643.58 M8926  
26202 0.0 0.0 0 0  
6 PAYMENT 7993.75 C760822614 11643.58 3649.83 M40514511 0  
.0 0.0 0  
6 PAYMENT 27202.42 C689471568 3649.83 0.0 M1058155740 0  
.0 0.0 0  
6 PAYMENT 18666.07 C1542792163 0.0 0.0 M148018969 0  
.0 0.0 0  
6 PAYMENT 20112.79 C1696963667 0.0 0.0 M1275013351 0  
.0 0.0 0  
6 PAYMENT 16419.57 C971497536 0.0 0.0 M630313530 0  
.0 0.0 0  
6 PAYMENT 5060.44 C1813976543 0.0 0.0 M1586868395 0.0 0  
.0 0  
6 PAYMENT 1031.66 C1484177113 0.0 0.0 M1943910383 0.0 0  
.0 0  
6 PAYMENT 7285.68 C1744369062 0.0 0.0 M337314074 0.0 0  
.0 0  
6 PAYMENT 1200.57 C2037835062 0.0 0.0 M1728726840 0.0 0  
.0 0  
Time taken: 0.085 seconds, Fetched: 5534 row(s)  
hive> Search for "TEXTFILE"
```

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we can execute several Hive Queries to understand the data and perform necessary analysis in Hive: The following are the queries used to extract insights from the queries :

1) Total Number of transactions by type :

```
query: SELECT type, AVG(amount) AS avg_transaction_amount FROM Fraud
GROUP BY type;
```

2) Money laundering transactions may involve multiple transfers with small amounts. You can identify such patterns by grouping transactions by nameOrig and calculating the total number of outgoing transactions and the total amount transferred for each account

```
SELECT nameOrig, COUNT(*) AS num_outgoing_transactions, SUM(amount)
AS total_outgoing_amount
```

```
FROM your_table_name
```

```
WHERE type IN ('TRANSFER', 'CASH_OUT')
```

```
GROUP BY nameOrig
```

```
HAVING num_outgoing_transactions > 1 AND total_outgoing_amount <1000;
```

3) Based on your fraud detection criteria, you can flag transactions that meet certain conditions as potentially fraudulent. For example, transactions where isFraud is not flagged but the amount is significantly higher than average.

```
query: SELECT F.*,
```

```
CASE
```

```
    WHEN F.isFraud = 0 AND F.amount > A.avg_amount THEN 1

    ELSE 0

END AS flagged_as_potential_fraud

FROM FraudTransactions F

CROSS JOIN (SELECT AVG(amount) as avg_amount FROM Fraud) A;
```

Insights:

- The last query will yield a new column which says potential fraud. This will flag transactions that meet certain criteria as potentially fraudulent.
- It will return all columns from the FraudTransactions table along with a new column flagged_as_potential_fraud. Transactions where isFraud is 0 (not flagged as fraud) and the amount is significantly higher than the average amount in the entire Fraud dataset will be flagged as 1; otherwise, they will be flagged as 0.
- Based on the isFraud and Potential_fraud transactions, there isn't much difference. So it's better to go with the same dataset we have taken.

Output :

cloudera-quickstart-vm-5.13.0-0-virtualbox [Running] - Oracle VM VirtualBox

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```
cloudera@quickstart:~  
File Edit View Search Terminal Help  
Time taken: 85.911 seconds, Fetched: 5534 row(s)  
hive> SELECT nameOrig,  
  > COUNT(*) AS num_outgoing_transactions,  
  > SUM(amount) AS total_outgoing_amount  
  > FROM Fraud  
  > WHERE type IN ('TRANSFER', 'CASH_OUT')  
  > GROUP BY nameOrig  
  > HAVING COUNT(*) > 1 AND SUM(amount) < 1000; -- Adjust the threshold as  
needed  
  >  
  >;  
Query ID = cloudera_20231127123434_f63cd7b2-6070-44d3-ad13-28e4167a17b5  
Total jobs = 1  
Launching Job 1 out of 1  
Number of reduce tasks not specified. Estimated from input data size: 1  
In order to change the average load for a reducer (in bytes):  
  set hive.exec.reducers.bytes.per.reducer=<number>  
In order to limit the maximum number of reducers:  
  set hive.exec.reducers.max=<number>  
In order to set a constant number of reducers:  
  set mapreduce.job.reduces=<number>  
Starting Job = job_1699823206635_0019, Search for "TEXTFILE" >:./quickstart.cloud
```

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cloudera-quickstart-vm-5.13.0-0-virtualbox [Running] - Oracle VM VirtualBox

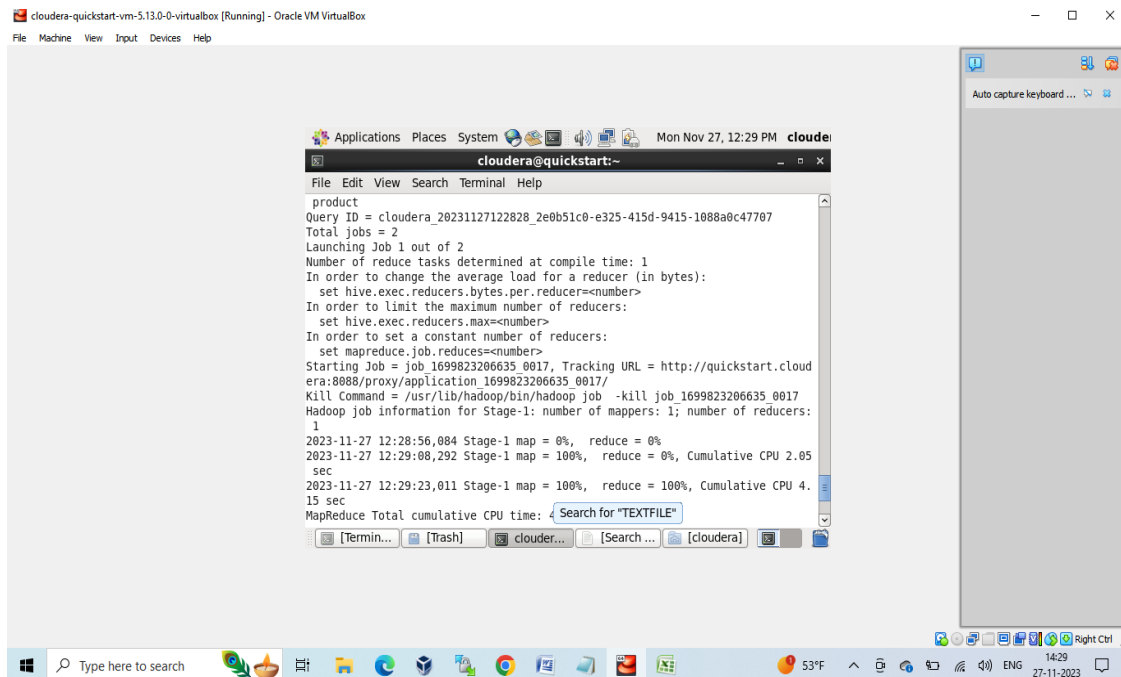
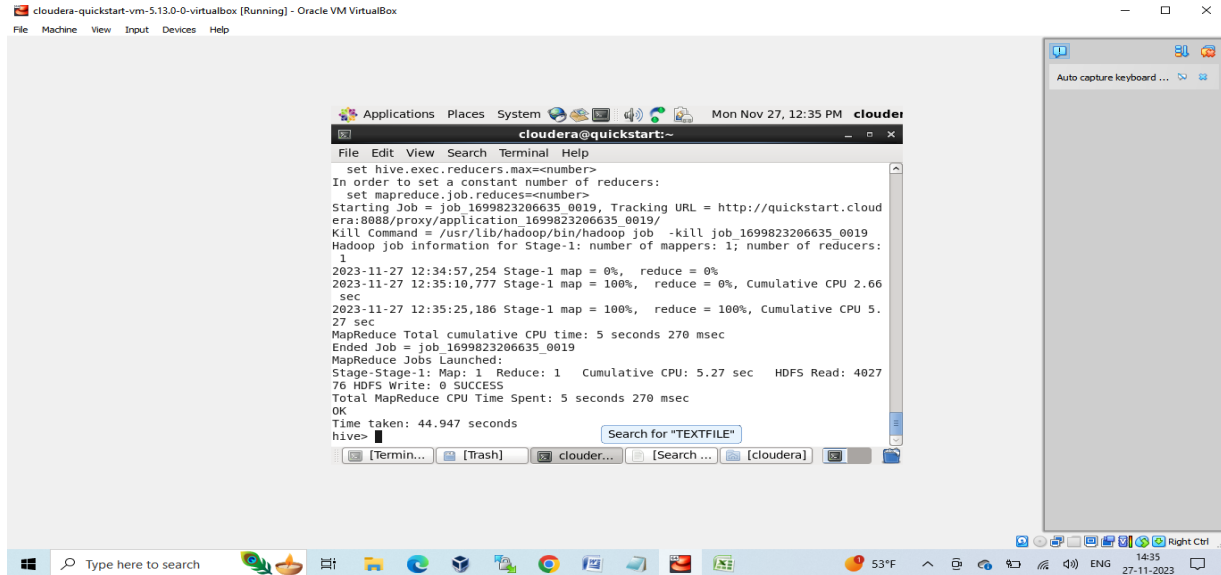
File Machine View Input Devices Help

Applications Places System Mon Nov 27, 12:22 PM cloudera

```
cloudera@quickstart:~  
File Edit View Search Terminal Help  
Hadoop job information for Stage-1: number of mappers: 1; number of reducers:  
1  
2023-11-27 12:20:13,850 Stage-1 map = 0%, reduce = 0%  
2023-11-27 12:20:30,266 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 2.79  
sec  
2023-11-27 12:20:46,434 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 5.  
26 sec  
MapReduce Total cumulative CPU time: 5 seconds 260 msec  
Ended Job = job_1699823206635_0012  
MapReduce Jobs Launched:  
Stage-Stage-1: Map: 1 Reduce: 1 Cumulative CPU: 5.26 sec HDFS Read: 4013  
40 HDFS Write: 139 SUCCESS  
Total MapReduce CPU Time Spent: 5 seconds 260 msec  
OK  
CASH IN 167517.1047992864  
CASH OUT 176662.91455344067  
DEBIT 3723.842035087719  
PAYMENT 6522.860295302012  
TRANSFER 441016.5432543102  
type NULL  
Time taken: 59.227 seconds, Fetched: 6_row(s)  
hive>
```

[Termin... [Trash] cloudera... [Search ... [cloudera]

Type here to search 53°F 14:22 27-11-2023



EMR Cluster Creation on EC2 instance (AWS) :

EC2 instance Launch :

Launch an instance [Info](#)

Amazon EC2 allows you to create virtual machines, or instances, that run on the AWS Cloud. Quickly get started by following the simple steps below.

Name and tags [Info](#)

Name
project1 [Add additional tags](#)

▼ Application and OS Images (Amazon Machine Image) [Info](#)

An AMI is a template that contains the software configuration (operating system, application server, and applications) required to launch your instance. Search or Browse for AMIs if you don't see what you are looking for below

▼ Summary

Number of instances [Info](#)
1

Software Image (AMI)
Amazon Linux 2023 AMI 2023.2.2...[read more](#)
ami-0250bd60aa48260c6

Virtual server type (instance type)
t2.micro

Firewall (security group)
New security group

Storage (volumes)
1 volume(s) - 8 GiB

[Free tier: In your first year](#) [X](#)

[Cancel](#) [Launch instance](#)

[Review commands](#)

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Cluster Creation :

The screenshot shows the 'Create cluster' page in the AWS Management Console. The page is titled 'Create cluster' and includes a 'Name and applications' section on the left and a 'Summary' section on the right. In the 'Name and applications' section, the cluster name is 'projectcluster', the Amazon EMR release is 'emr-6.13.0', and the application bundle is 'Custom'. The 'Custom' bundle includes Hadoop 3.3.3, Hive 3.1.3, and Presto 3.1.3. The 'Summary' section shows the cluster configuration, including the instance groups: Primary (m5.xlarge), Core (m5.xlarge), and Task (m5.xlarge).

Name and applications

Name: projectcluster

Amazon EMR release: emr-6.13.0

Application bundle: Custom (Hadoop 3.3.3, Hive 3.1.3, Hue 4.11.0, JupyterEnterpriseGateway 2.6.0, JupyterHub...)

Cluster configuration

Instance groups: Primary (m5.xlarge), Core (m5.xlarge), Task (m5.xlarge)

Workspace Creation :

The screenshot shows the 'EMR Studio: Studios' page in the AWS Management Console. The page displays a list of studios, including 'Studio 1' and 'Studio 2'. The 'Create Studio' button is visible. The page also includes a sidebar with navigation options for Amazon EMR, EMR on EC2, EMR on EKS, and EMR Studio.

Studios (2)

Studio name	Creation time (UTC-06:00)	Authenticated by	Studio Access URL
Studio 2	November 28, 2023, 11:05	IAM	https://es-2EI9VHFQROUXEDPFU50...
Studio 1	November 28, 2023, 11:01	IAM	https://es-DAWZOWTFEMURFR9PDRI...

Loading the Notebook for writing the code in pyspark :

The screenshot displays the Amazon EMR Studio interface. On the left, the 'Compute' sidebar shows options for attaching the workspace to a compute layer. The 'Compute type' section is expanded, showing 'EMR on EC2 cluster' as the selected option. The 'Runtime role' section indicates that a runtime role must be chosen for Amazon EMR. The main area shows a 'Spark Notebook (2).ipynb' with the following code:

```
[16]: pip install pyspark
      sc.install_pypi_package("pandas==1.0.7") # Install a specific version of pandas
      sc.install_pypi_package("matplotlib", "https://pypi.org/simple") # Install the latest version of matplotlib from the given PyPI
      sc.install_pypi_package("seaborn", "https://pypi.org/simple")
      #sc.install_pypi_package("sklearn", "https://github.com/scikit-learn/skLearn-pypi-package")
      sc.install_pypi_package("scikit-learn")

[ ]: #Importing Libraries
     from pyspark.ml.classification import RandomForestClassifier
     from pyspark.ml.evaluation import BinaryClassificationEvaluator
     from pyspark import SparkContext
     from pyspark import SparkConf
     from pyspark.sql import SparkSession
     from pyspark.sql.functions import lit
     from pyspark.ml.feature import VectorAssembler, StandardScaler
     from pyspark.ml import Pipeline
     #For visualizations
     import seaborn as sns
     import matplotlib.pyplot as plt
     from sklearn.metrics import roc_curve, auc

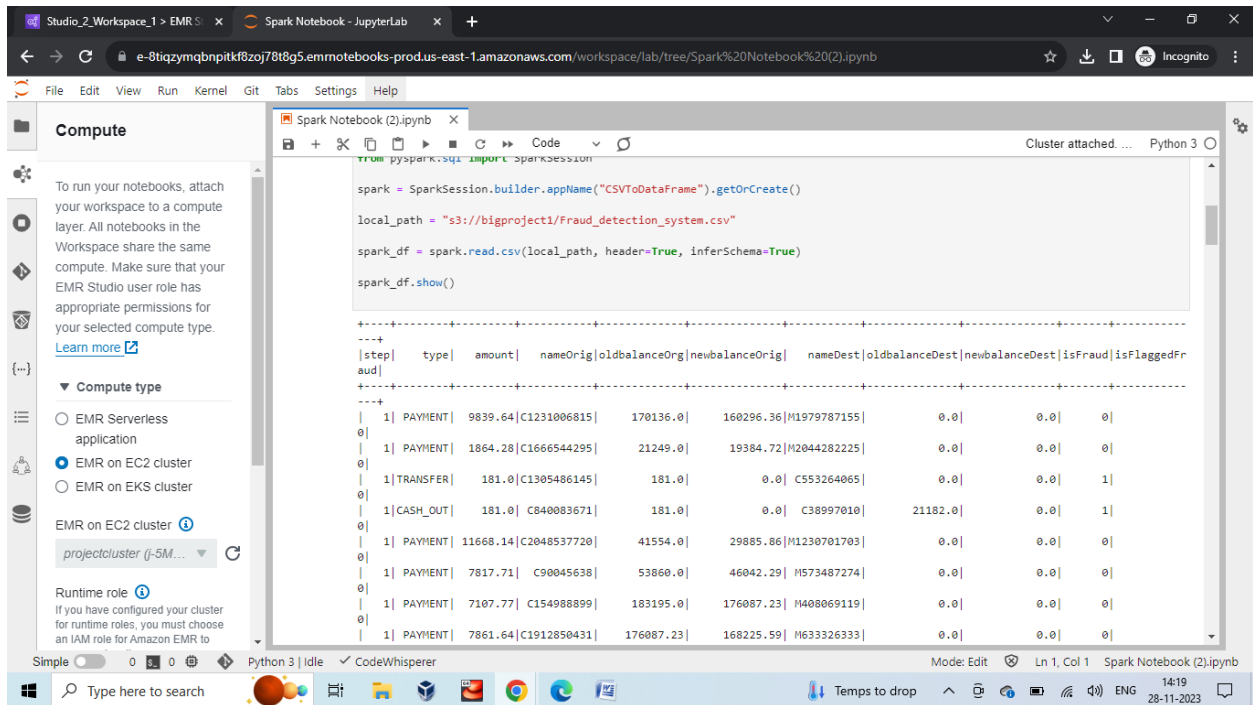
[17]: #Loading the dataset into Spark's Data frame
     from pyspark.sql import SparkSession

     spark = SparkSession.builder.appName("CSVToDataFrame").getOrCreate()

     local_path = "s3://bigproject1/Fraud_detection_system.csv"
```

The bottom status bar shows the interface is in 'Simple' mode, using 'Python 3' and 'CodeWhisperer'. The system tray at the bottom indicates a temperature of 61°F, sunny weather, and the date 28-11-2023.

Loading the data into Spark Data Frame :



The screenshot shows the Amazon EMR Studio interface. On the left, the 'Compute' sidebar is visible, showing the 'EMR on EC2 cluster' selected. The main area displays a Spark Notebook (2).ipynb with the following code:

```
from pyspark.sql import SparkSession

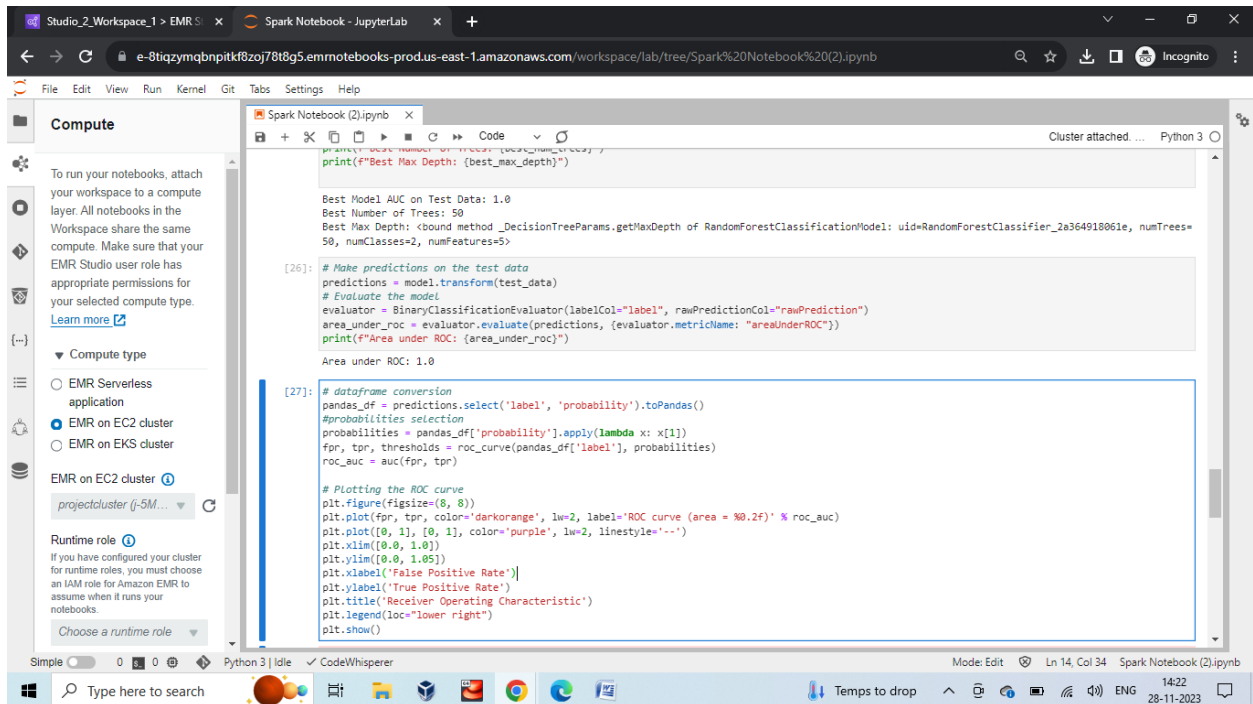
spark = SparkSession.builder.appName("CSVtoDataFrame").getOrCreate()

local_path = "s3://bigproject1/Fraud_detection_system.csv"

spark_df = spark.read.csv(local_path, header=True, inferSchema=True)

spark_df.show()
```

The output of the code is a table with 10 columns: `step`, `type`, `amount`, `nameOrig`, `oldbalanceOrig`, `newbalanceOrig`, `nameDest`, `oldbalanceDest`, `newbalanceDest`, `isFraud`, and `isFlaggedFraud`. The table contains 10 rows of data, including transactions like 'PAYMENT', 'TRANSFER', and 'CASH_OUT'.



The screenshot shows the Amazon EMR Studio interface. On the left, the 'Compute' sidebar is visible, showing the 'EMR on EC2 cluster' selected. The main area displays a Spark Notebook (2).ipynb with the following code:

```
print("Best Model AUC on Test Data: 1.0")
print("Best Number of Trees: 50")
print("Best Max Depth: <bound method _DecisionTreeParams.getMaxDepth of RandomForestClassificationModel: uid=RandomForestClassifier_2a364918061e, numTrees=50, numClasses=2, numFeatures=5>")

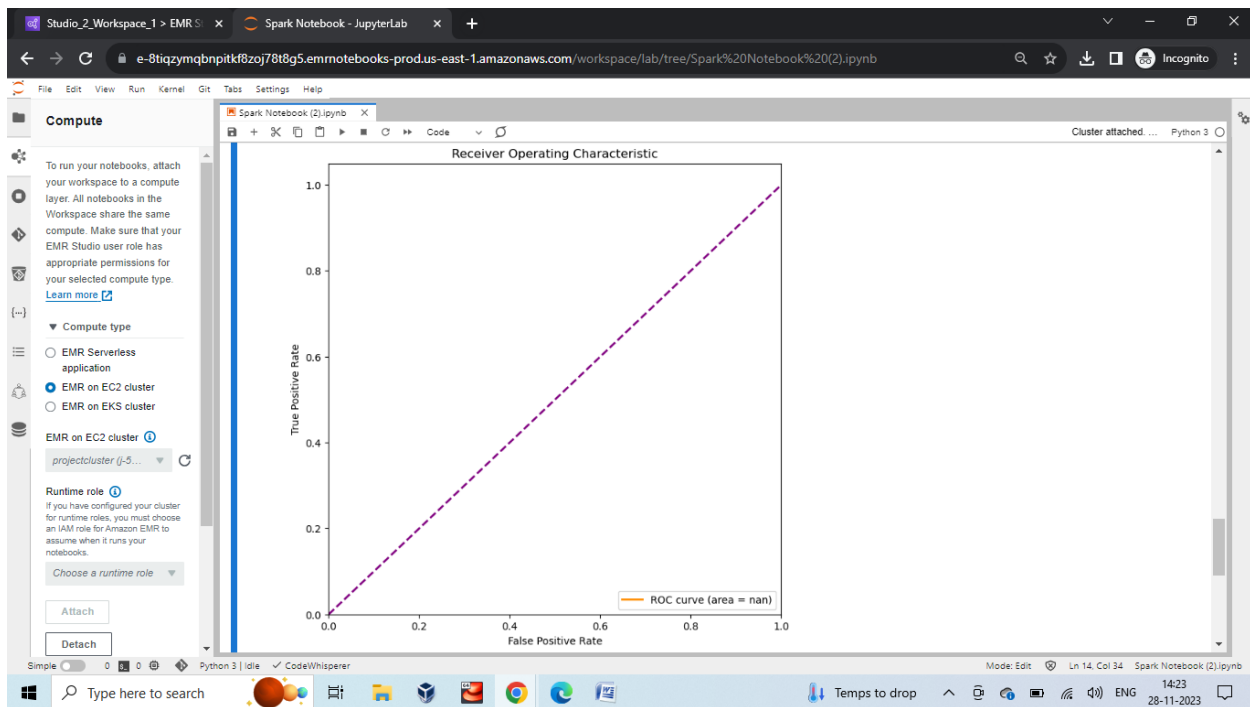
[26]: # Make predictions on the test data
      predictions = model.transform(test_data)
      # Evaluate the model
      evaluator = BinaryClassificationEvaluator(labelCol="label", rawPredictionCol="rawPrediction")
      area_under_roc = evaluator.evaluate(predictions, {evaluator.metricName: "areaUnderROC"})
      print(f"Area under ROC: {area_under_roc}")

      Area under ROC: 1.0

[27]: # dataframe conversion
      pandas_df = predictions.select('label', 'probability').toPandas()
      #probabilities selection
      probabilities = pandas_df['probability'].apply(lambda x: x[1])
      fpr, tpr, thresholds = roc_curve(pandas_df['label'], probabilities)
      roc_auc = auc(fpr, tpr)

      # Plotting the ROC curve
      plt.figure(figsize=(8, 8))
      plt.plot(fpr, tpr, color='darkorange', lw=2, label='ROC curve (area = %.2f)' % roc_auc)
      plt.plot([0, 1], [0, 1], color='purple', lw=2, linestyle='--')
      plt.xlim([0.0, 1.0])
      plt.ylim([0.0, 1.05])
      plt.xlabel('False Positive Rate')
      plt.ylabel('True Positive Rate')
      plt.title('Receiver Operating Characteristic')
      plt.legend(loc='lower right')
      plt.show()
```

The code evaluates a machine learning model and plots an ROC curve. The output shows the 'Area under ROC' as 1.0. The ROC curve plot is also displayed, showing a dark orange line representing the ROC curve and a purple dashed line representing the random baseline.



- **Final Conclusion:**

- The Receiver Operating Characteristic (ROC) curve offers a graphical depiction of how well a model can differentiate between two categories, such as positive and negative instances. This curve enables the evaluation of the model's proficiency in distinguishing true positives (accurately identified positive cases) from false positives (wrongly identified positive cases), over a range of probability thresholds.
- The area under ROC curve indicates that 1 which indicates a perfect classifier
- The “evaluate method” of the evaluator computes the metric specified for the model predictions provided.
- In this case, {evaluator.metricName: "areaUnderROC"} sets the metric to the Area Under the ROC Curve (AUC - ROC).

- The AUC - ROC is a single scalar value that summarizes the performance of the binary classification model across all classification thresholds.
- It essentially measures the ability of the model to discriminate between the two classes.