from google.colab import drive
drive.mount('/content/drive')

Ery Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force\_remount=True).

import numpy as np
import pandas as pd
import seaborn as sns

from sklearn.model\_selection import train\_test\_split, cross\_val\_score

from sklearn.linear\_model import LinearRegression
from sklearn.metrics import mean\_squared\_error

from sklearn.metrics import confusion\_matrix

from sklearn.model\_selection import train\_test\_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import classification\_report, confusion\_matrix
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import classification\_report, confusion\_matrix, accuracy\_score
import matplotlib.pyplot as plt
import seaborn as sns

df=pd.read\_csv("//content/drive/MyDrive/archive (15).zip")

df

<b>→</b>		Index	Incoming Calls	Answered Calls	Answer Rate	Abandoned Calls	Answer Speed (AVG)	Talk Duration (AVG)	Waiting Time (AVG)	Service Level (20 Seconds)
	0	1	217	204	94.01%	13	0:00:17	0:02:14	0:02:45	76.28%
	1	2	200	182	91.00%	18	0:00:20	0:02:22	0:06:55	72.73%
	2	3	216	198	91.67%	18	0:00:18	0:02:38	0:03:50	74.30%
	3	4	155	145	93.55%	10	0:00:15	0:02:29	0:03:12	79.61%
	4	5	37	37	100.00%	0	0:00:03	0:02:06	0:00:35	97.30%
	1246	1247	191	184	96.34%	7	0:00:07	0:02:50	0:01:56	92.55%
	1247	1248	212	209	98.58%	3	0:00:10	0:02:51	0:01:45	89.10%
	1248	1249	210	203	96.67%	7	0:00:12	0:03:22	0:03:52	85.24%
	1249	1250	167	159	95.21%	8	0:00:16	0:03:16	0:02:42	83.03%
	1250	1251	12	11	91.67%	1	0:00:03	0:02:12	0:00:58	91.67%

1251 rows × 9 columns

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1251 entries, 0 to 1250
Data columns (total 9 columns):

#	Column	Non-Null Count	Dtype
0	Index	1251 non-null	int64
1	Incoming Calls	1251 non-null	int64
2	Answered Calls	1251 non-null	int64
3	Answer Rate	1251 non-null	object
4	Abandoned Calls	1251 non-null	int64
5	Answer Speed (AVG)	1251 non-null	object
6	Talk Duration (AVG)	1251 non-null	object
7	Waiting Time (AVG)	1251 non-null	object
8	Service Level (20 Seconds)	1251 non-null	object

dtypes: int64(4), object(5)
memory usage: 88.1+ KB

df.isnull().sum()

```
₹
               Index
           Incoming Calls
           Answered Calls
            Answer Rate
          Abandoned Calls
         Answer Speed (AVG)
                                 0
         Talk Duration (AVG)
         Waiting Time (AVG)
                                 0
      Service Level (20 Seconds) 0
     dtype: int64
df.dropna(inplace=True)
X = df[['Incoming Calls', 'Answered Calls']]
y = df['Answer Rate']
```

X\_train, X\_test, y\_train, y\_test = train\_test\_split(X, y, test\_size=0.2, random\_state=42)

scaler = StandardScaler()
X\_train = scaler.fit\_transform(X\_train)
X\_test = scaler.transform(X\_test)

model = RandomForestClassifier(random\_state=42)

model.fit(X\_train, y\_train)

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RandomForestClassifier ① ②

RandomForestClassifier(random\_state=42)

y\_pred = model.predict(X\_test)
print("\nClassification Report:")
print(classification\_report(y\_test, y\_pred))

Classification Report: precision recall f1-score support 100.00% 0.85 0.52 0.65 21 51.72% 0.00 0.00 0.00 0 53.46% 0.00 0.00 0.00 0 56.29% 0.00 0.00 0.00 1 56.95% 0.00 0.00 0.00 1 67.49% 0.00 0.00 0.00 1 72.97% 0.00 0.00 0.00 1 73.77% 0.00 0.00 0.00 1 74.08% 0.00 0.00 0.00 74.71% 0.00 0.00 0.00 1 75.00% 0.00 0.00 0.00 0 77.08% 0.00 0.00 0.00 77.84% 0.00 0.00 1 0.00 77.92% 0.00 0.00 0.00 0 78.95% 0.00 0.00 0.00 0 79.07% 0.00 0.00 0.00 1 79.31% 0.00 0.00 0.00 1 79.60% 0.00 0.00 0.00 1

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```

```
accuracy = accuracy_score(y_test, y_pred)
accuracy
0.18326693227091634
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)
LABELS = ['Normal', 'Fraud']
print("Confusion Matrix:")
print(cm)
plt.figure(figsize=(10, 6))
sns.heatmap(cm, annot=True, fmt="d", cmap='Blues', xticklabels=LABELS, yticklabels=LABELS)
plt.title("Confusion Matrix")
plt.ylabel("True Class")
plt.xlabel("Predicted Class")
plt.show()
```

```
Confusion Matrix:

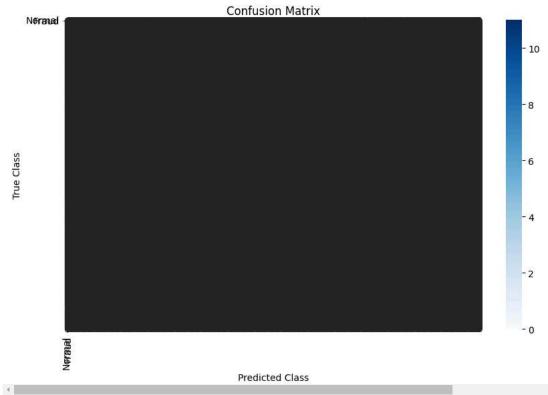
[[11 0 0 ... 1 0 1]

[ 0 0 0 ... 0 0 0]

[ 0 0 0 ... 0 0 0]

[ 0 0 0 ... 0 0 0]

[ 0 0 0 ... 0 0 0]
```



```
new_call = [[10, 5]]
new_call_scaled = scaler.transform(new_call)
prediction = model.predict(new_call_scaled)

if prediction[0] == 1:
    print("The call is predicted to be fraud.")
```