

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from wordcloud import WordCloud
import nltk
import re
from nltk.stem.porter import PorterStemmer
from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC
from sklearn.naive_bayes import MultinomialNB
```

```
[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data]   Unzipping corpora/stopwords.zip.
```

```
df = pd.read_csv('/content/drive/My Drive/Subway/fraudTrain.csv')
```

```
df.columns
```

```
Index(['Unnamed: 0', 'trans_date_trans_time', 'cc_num', 'merchant', 'category',
      'amt', 'first', 'last', 'gender', 'street', 'city', 'state', 'zip',
      'lat', 'long', 'city_pop', 'job', 'dob', 'trans_num', 'unix_time',
      'merch_lat', 'merch_long', 'is_fraud'],
      dtype='object')
```

```
df.shape
```

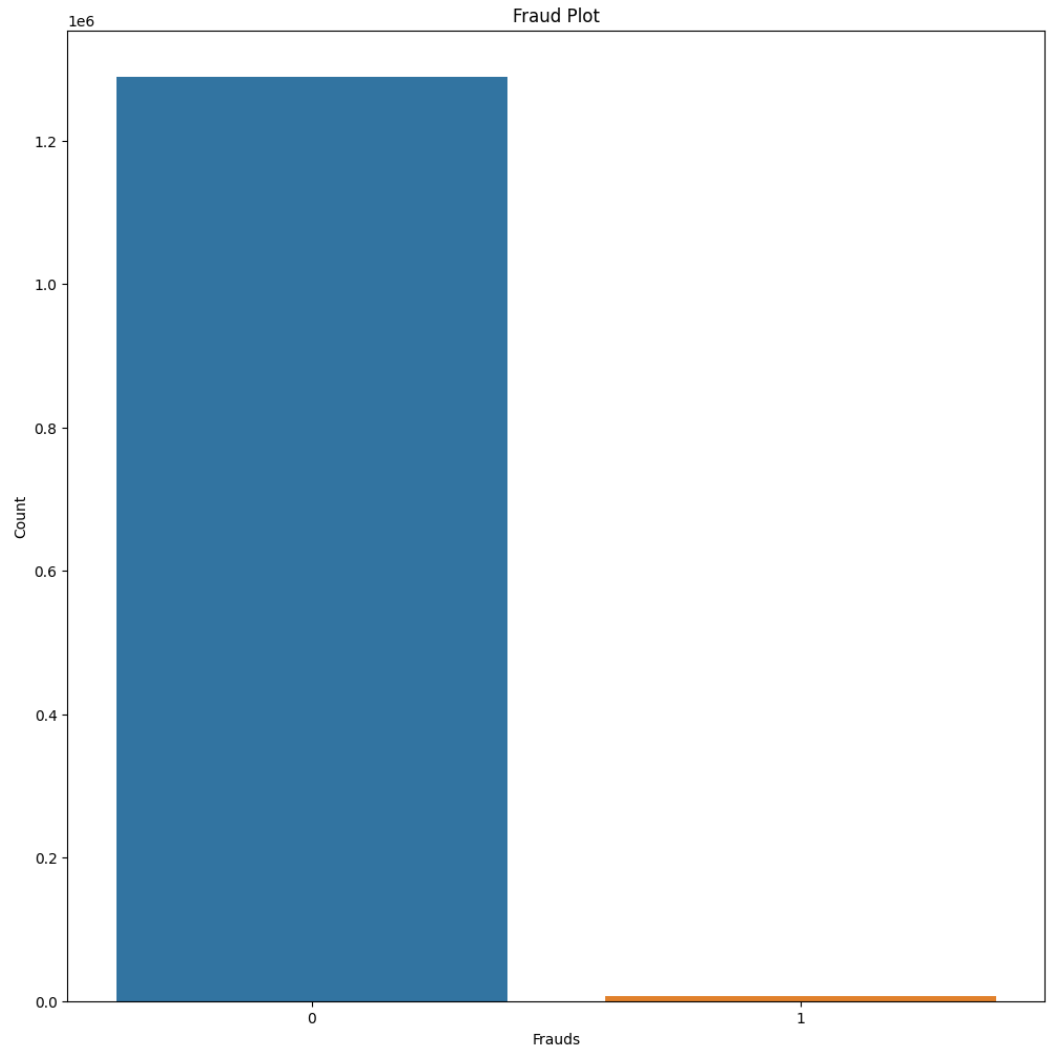
```
(1296675, 23)
```

```
df.head(5)
```

	Unnamed: 0	trans_date_trans_time	cc_num	merchant	category	amt	first	last
0	0	2019-01-01 00:00:18	2703186189652095	fraud_Rippin, Kub and Mann	misc_net	4.97	Jennifer	Ba
1	1	2019-01-01 00:00:44	630423337322	fraud_Heller, Gutmann and Zieme	grocery_pos	107.23	Stephanie	
2	2	2019-01-01 00:00:51	38859492057661	fraud_Lind-Buckridge	entertainment	220.11	Edward	Sanc
3	3	2019-01-01 00:01:16	3534093764340240	fraud_Kutch, Hermiston and Farrell	gas_transport	45.00	Jeremy	W
4	4	2019-01-01 00:03:06	375534208663984	fraud_Keeling-Crist	misc_pos	41.96	Tyler	Ga

```
5 rows x 23 columns
```

```
plt.figure(figsize=(12,12))
sns.countplot(x='is_fraud', data=df)
plt.xlabel('Frauds')
plt.ylabel('Count')
plt.title('Fraud Plot')
plt.show()
```



```
frauds = list(df['is_fraud'].unique())
frauds.sort()
frauds
```

[0, 1]

```
df.isna().any()

Unnamed: 0      False
trans_date_trans_time  False
cc_num          False
merchant        False
category        False
amt             False
first           False
last            False
gender          False
street          False
city            False
state           False
zip             False
lat             False
long            False
city_pop        False
job             False
dob             False
trans_num       False
unix_time       False
merch_lat       False
merch_long      False
is_fraud        False
dtype: bool
```

```
df.drop('Unnamed: 0', axis=1, inplace=True)
```

```
import pandas as pd
from sklearn.preprocessing import OrdinalEncoder
```

```

from sklearn.naive_bayes import CategoricalNB
from sklearn import metrics

data = df
le = OrdinalEncoder()
data[['trans_date_trans_time', 'cc_num', 'merchant', 'category',
      'amt', 'first', 'last', 'gender', 'street', 'city', 'state', 'zip',
      'lat', 'long', 'city_pop', 'job', 'dob', 'trans_num', 'unix_time',
      'merch_lat', 'merch_long', 'is_fraud']] = le.fit_transform(
data[['trans_date_trans_time', 'cc_num', 'merchant', 'category',
      'amt', 'first', 'last', 'gender', 'street', 'city', 'state', 'zip',
      'lat', 'long', 'city_pop', 'job', 'dob', 'trans_num', 'unix_time',
      'merch_lat', 'merch_long', 'is_fraud']])

Features = ['trans_date_trans_time', 'cc_num', 'merchant', 'category',
            'amt', 'first', 'last', 'gender', 'street', 'city', 'state', 'zip',
            'lat', 'long', 'city_pop', 'job', 'dob', 'trans_num', 'unix_time',
            'merch_lat', 'merch_long']
X = data[Features]
Y = data['is_fraud']

clf_nb = CategoricalNB()
clf_nb.fit(X,Y)

# y_pred = clf.predict(X)
# print(Y)
# print(y_pred)

# print("Accuracy : " , metrics.accuracy_score(y_pred,Y) * 100)

```

▼ CategoricalNB
CategoricalNB()

```

clf_lr = LogisticRegression(max_iter=1000)
clf_lr.fit(X, Y)

# y_pred = clf_lr.predict(X)
# print(Y)
# print(y_pred)

# print("Accuracy : ", metrics.accuracy_score(y_pred, Y) * 100)

```

▼ LogisticRegression
LogisticRegression(max_iter=1000)

```

from sklearn.svm import SVC

clf_svm = SVC(kernel='linear')
clf_svm.fit(X, Y)

y_pred = clf_nb.predict(X)
print(Y)
print(y_pred)

print("Accuracy : " , metrics.accuracy_score(y_pred,Y) * 100)

```

```

0      0.0
1      0.0
2      0.0
3      0.0
4      0.0
...
1296670 0.0
1296671 0.0
1296672 0.0
1296673 0.0
1296674 0.0
Name: is_fraud, Length: 1296675, dtype: float64
[0. 0. 0. ... 0. 0. 0.]
Accuracy : 92.29274876125476

```

```

y_pred = clf_lr.predict(X)
print(Y)
print(y_pred)

print("Accuracy : " , metrics.accuracy_score(y_pred,Y) * 100)

```

```

0      0.0
1      0.0
2      0.0
3      0.0
4      0.0
...
1296670 0.0
1296671 0.0
1296672 0.0
1296673 0.0
1296674 0.0
Name: is_fraud, Length: 1296675, dtype: float64
[0. 0. 0. ... 0. 0. 0.]
Accuracy : 99.39792160718761

```

```

y_pred = clf_svm.predict(X)
print(Y)
print(y_pred)

```

```
print("Accuracy : " , metrics.accuracy_score(y_pred,Y) * 100)
```

```
tt = pd.read_csv('/content/drive/My Drive/Subway/fraudTest.csv')
```

```

tt[['trans_date_trans_time', 'cc_num', 'merchant', 'category',
    'amt', 'first', 'last', 'gender', 'street', 'city', 'state', 'zip',
    'lat', 'long', 'city_pop', 'job', 'dob', 'trans_num', 'unix_time',
    'merch_lat', 'merch_long', 'is_fraud']] = le.fit_transform(
tt[['trans_date_trans_time', 'cc_num', 'merchant', 'category',
    'amt', 'first', 'last', 'gender', 'street', 'city', 'state', 'zip',
    'lat', 'long', 'city_pop', 'job', 'dob', 'trans_num', 'unix_time',
    'merch_lat', 'merch_long', 'is_fraud']])

```

```

XX= tt[Features]
YY = tt['is_fraud']

```

```

yy_pred = clf_nb.predict(XX)
print(YY)
print(yy_pred)

```

```
print("Accuracy : " , metrics.accuracy_score(yy_pred,YY) * 100)
```

```

0      0.0
1      0.0
2      0.0
3      0.0
4      0.0
...
555714 0.0
555715 0.0
555716 0.0
555717 0.0
555718 0.0
Name: is_fraud, Length: 555719, dtype: float64
[0. 0. 0. ... 0. 0. 0.]
Accuracy : 72.92966409282388

```

```

yy_pred = clf_lr.predict(XX)
print(YY)
print(yy_pred)

```

```
print("Accuracy : " , metrics.accuracy_score(yy_pred,YY) * 100)
```

```

0      0.0
1      0.0
2      0.0
3      0.0
4      0.0
...
555714 0.0
555715 0.0
555716 0.0
555717 0.0
555718 0.0
Name: is_fraud, Length: 555719, dtype: float64
[0. 0. 0. ... 0. 0. 0.]
Accuracy : 99.4983075979047

```

```

yy_pred = clf_svm.predict(XX)
print(YY)
print(yy_pred)

```

```
print("Accuracy : " , metrics.accuracy_score(yy_pred,YY) * 100)
```

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