```
import numpy as np
import pandas as pd
import os
for dirname, _, filenames in os.walk('onlinefraud.csv'):
    for filename in filenames:
       print(os.path.join(dirname, filename))
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
from sklearn.model_selection import train_test_split
from xgboost import XGBClassifier
from \ sklearn.linear\_model \ import \ LogisticRegression
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import confusion_matrix, recall_score, f1_score
import scipy.stats as st
import warnings
warnings.filterwarnings('ignore')
plt.style.use('fivethirtyeight')
pd.set_option('display.max_columns', None)
pd.set_option('display.max_rows', None)
df = pd.read_csv("onlinefraud.csv")
df.head()
```

nameDes	newbalanceOrig	oldbalanceOrg	nameOrig	amount	type	step	
M197978715	160296.36	170136.0	C1231006815	9839.64	PAYMENT	1	0
M204428222	19384.72	21249.0	C1666544295	1864.28	PAYMENT	1	1
C55326406	0.00	181.0	C1305486145	181.00	TRANSFER	1	2
C3899701	0.00	181.0	C840083671	181.00	CASH_OUT	1	3
M123070170	29885.86	41554.0	C2048537720	11668.14	PAYMENT	1	4

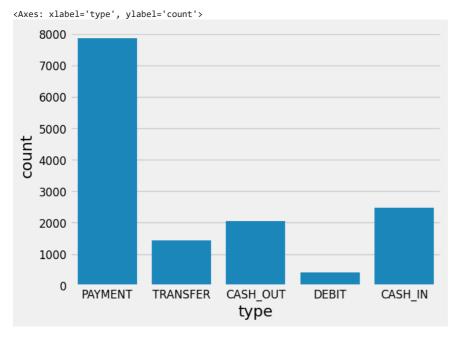
df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14247 entries, 0 to 14246
Data columns (total 11 columns):
# Column
                   Non-Null Count Dtype
                    14247 non-null int64
    step
                     14247 non-null object
    type
                    14247 non-null float64
 2
    amount
    nameOrig
                     14247 non-null object
 3
    nameorig 14247 non-null object oldbalanceOrg 14246 non-null float64
     newbalanceOrig 14246 non-null float64
    nameDest
                     14246 non-null object
    oldbalanceDest 14246 non-null float64
    newbalanceDest 14246 non-null float64
9 isFraud 14246 non-null float64
10 isFlaggedFraud 14246 non-null float64
dtypes: float64(7), int64(1), object(3)
memory usage: 1.2+ MB
```

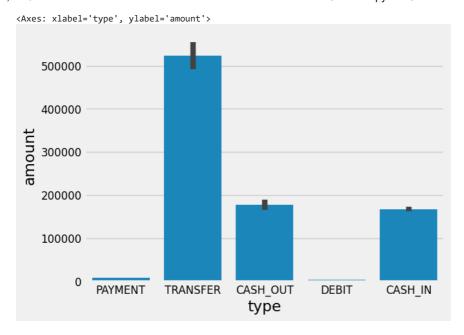
df.describe()

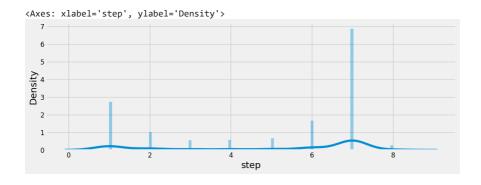
	step	amount	oldbalanceOrg	newbalanceOrig	oldbalanceDest	newt
count	14247.000000	1.424700e+04	1.424600e+04	1.424600e+04	1.424600e+04	1
mean	5.037131	1.118848e+05	7.937369e+05	8.107136e+05	8.411336e+05	1
std	2.463112	2.805152e+05	2.016693e+06	2.059984e+06	2.528174e+06	3
min	1.000000	2.390000e+00	0.000000e+00	0.000000e+00	0.000000e+00	0
25%	2.000000	4.585130e+03	0.000000e+00	0.000000e+00	0.000000e+00	0
50%	6.000000	1.286936e+04	2.041500e+04	8.216325e+03	0.000000e+00	0
75%	7.000000	1.218917e+05	1.384956e+05	1.294173e+05	2.641427e+05	2
max	8.000000	1.000000e+07	1.293042e+07	1.301050e+07	2.093759e+07	2

sns.countplot(x='type', data=df)



sns.barplot(x='type', y='amount', data=df)





```
numeric_df = df.select_dtypes(include=['number'])
plt.figure(figsize=(12, 6))
sns.heatmap(numeric_df.corr(), cmap='BrBG', fmt='.2f', linewidths=2, annot=True)
```



type_new = pd.get_dummies(df['type'], drop_first=True)
df_new = pd.concat([df, type_new], axis=1)
df_new.head()

	step	type	amount	nameOrig	oldbalanceOrg	newbalanceOrig	nameDes
0	1	PAYMENT	9839.64	C1231006815	170136.0	160296.36	M197978715
1	1	PAYMENT	1864.28	C1666544295	21249.0	19384.72	M204428222
2	1	TRANSFER	181.00	C1305486145	181.0	0.00	C55326406
3	1	CASH_OUT	181.00	C840083671	181.0	0.00	C3899701
4	1	PAYMENT	11668.14	C2048537720	41554.0	29885.86	M123070170

```
X = df.drop(['isFraud', 'type', 'nameOrig', 'nameDest'], axis=1)
y = df['isFraud']
X.shape, y.shape
     ((545326, 7), (545326,))
 X\_train, \ X\_test, \ y\_train, \ y\_test = train\_test\_split(X, \ y, \ test\_size=0.30, \ random\_state=42) 
#Train Test split - By default train_test_split does STRATIFIED split based on label (y-value).
x\_train, x\_test, y\_train, y\_test = train\_test\_split(X,Y,test\_size=0.2, random\_state=0)
XGB = XGBClassifier()
XGB.fit(X_train, y_train)
XGB.score(X_train,y_train)
     0.9999083116774248
\# Check for NaN values in y_test
nan_indices = np.isnan(y_test)
# Print the indices of NaN values
print(np.where(nan_indices))
# Drop rows with NaN values in y_test
X_test = X_test[~nan_indices]
y_test = y_test[~nan_indices]
     (array([50370]),)
XGB.score(X_test,y_test)
```

```
0.999724933831305
LR = LogisticRegression()
LR.fit(X_train,y_train)
LR.score(X_train,y_train)
     0.9996044303797469
LR.score(X_test,y_test)
     0.9996760331790925
from sklearn.impute import SimpleImputer
#Removing NaN rows
nan_indices = np.isnan(y)
X = X[~nan_indices]
y = y[~nan_indices]
#Imputing NaN values
imputer = SimpleImputer(strategy='mean')
y = imputer.fit_transform(y.values.reshape(-1, 1)).flatten()
\ensuremath{\mathtt{\#}} 
 Now you can proceed with predicting and constructing the confusion matrix
y_pred = XGB.predict(X)
cm = confusion_matrix(y, y_pred)
plt.figure(figsize=(8, 6))
sns.heatmap(cm, annot=True, fmt="d", cmap="Blues", cbar=False)
plt.xlabel('Predicted Labels')
plt.ylabel('True Labels')
plt.title('Confusion Matrix')
plt.show()
```

