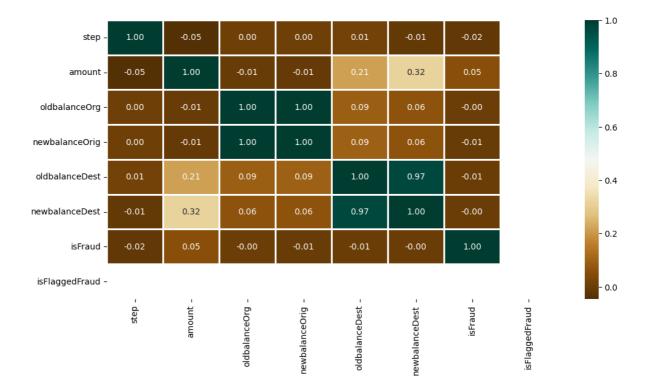
```
#Importing the required libraries
In [1]:
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
In [2]:
        #read csv file
         data = pd.read_csv('onlinefraud.csv')
         data.head()
Out[2]:
                                     nameOrig oldbalanceOrg newbalanceOrig
           step
                                                                              nameDest oldbala
                     type
                           amount
        0
              1
                 PAYMENT
                           9839.64 C1231006815
                                                    170136.0
                                                                  160296.36 M1979787155
        1
                 PAYMENT
                           1864.28 C1666544295
                                                     21249.0
                                                                   19384.72 M2044282225
        2
                TRANSFER
                            181.00 C1305486145
                                                       181.0
                                                                       0.00
                                                                             C553264065
        3
              1 CASH_OUT
                                                                       0.00
                            181.00
                                    C840083671
                                                       181.0
                                                                              C38997010
                                                                   29885.86 M1230701703
        4
                 PAYMENT 11668.14 C2048537720
                                                     41554.0
         data.shape
In [3]:
        (525391, 11)
Out[3]:
In [4]:
        data.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 525391 entries, 0 to 525390
        Data columns (total 11 columns):
         #
             Column
                             Non-Null Count
                                               Dtype
             ----
         ---
                              -----
         0
                              525391 non-null
                                               int64
             step
         1
                              525391 non-null object
             type
         2
             amount
                            525391 non-null float64
         3
             nameOrig
                            525391 non-null object
         4
             oldbalanceOrg 525391 non-null float64
             newbalanceOrig 525391 non-null float64
         5
         6
             nameDest
                              525391 non-null object
         7
             oldbalanceDest 525391 non-null float64
         8
             newbalanceDest 525391 non-null float64
         9
             isFraud
                              525391 non-null int64
         10 isFlaggedFraud 525391 non-null int64
        dtypes: float64(5), int64(3), object(3)
        memory usage: 44.1+ MB
In [5]:
         data.isna().sum()
                           0
        step
Out[5]:
                           0
        type
        amount
                           0
        nameOrig
                           0
        oldbalanceOrg
                           0
        newbalanceOrig
                           0
        nameDest
                           0
                           0
        oldbalanceDest
        newbalanceDest
                           0
        isFraud
                           0
        isFlaggedFraud
                           0
        dtype: int64
```

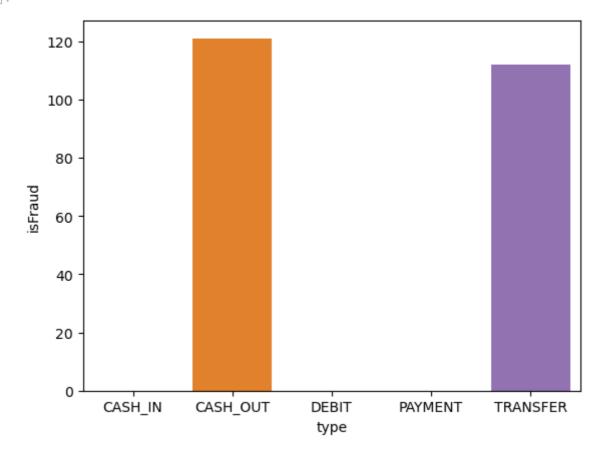
```
data.describe()
In [6]:
                                   amount oldbalanceOrg newbalanceOrig oldbalanceDest newbalanceI
Out[6]:
                         step
                                                                           5.253910e+05
          count 525391.000000 5.253910e+05
                                             5.253910e+05
                                                             5.253910e+05
                                                                                           5.253910e
          mean
                     14.229581
                              1.645037e+05
                                             9.038810e+05
                                                             9.235996e+05
                                                                           9.817080e+05
                                                                                           1.156670e
                                                             3.039871e+06
            std
                     3.851198 2.696783e+05
                                             3.002894e+06
                                                                           2.331355e+06
                                                                                           2.499715e
                      1.000000
                               1.000000e-01
                                             0.000000e+00
                                                             0.000000e+00
                                                                           0.000000e+00
                                                                                           0.000000e
            min
           25%
                                             0.000000e+00
                                                             0.000000e+00
                                                                           0.000000e+00
                                                                                           0.000000e
                     11.000000 1.315852e+04
           50%
                     15.000000 8.034980e+04
                                             1.840700e+04
                                                             0.000000e+00
                                                                           1.211080e+05
                                                                                           2.213392e
           75%
                     18.000000 2.198746e+05
                                             1.663035e+05
                                                             2.059076e+05
                                                                           8.981122e+05
                                                                                           1.195243e
                     20.000000 1.000000e+07
                                             3.890000e+07
                                                             3.890000e+07
                                                                           4.150000e+07
                                                                                           4.150000e
           max
 In [7]:
          #Understanding the transaction type
          print(data.type.value_counts())
          CASH_OUT
                      191642
          PAYMENT
                      172514
          CASH IN
                       114831
          TRANSFER
                       42603
          DEBIT
                         3801
          Name: type, dtype: int64
In [8]: #Count the occurrences of fraud and no fraud and print them
          occ = data['isFraud'].value_counts()
          occ
          0
               525158
Out[8]:
          1
                  233
          Name: isFraud, dtype: int64
In [9]:
         #Print the ratio of fraud cases
          ratio_cases = occ/len(data.index)
          print(f'Ratio of fraudulent cases: {ratio_cases[1]}\nRatio of non-fraudulent cases:
          Ratio of fraudulent cases: 0.00044347923736797926
          Ratio of non-fraudulent cases: 0.999556520762632
          plt.figure(figsize=(12, 6))
In [10]:
          sns.heatmap(data.corr(),
                       cmap='BrBG',
                       fmt='.2f',
                       linewidths=2,
                       annot=True)
          <AxesSubplot:>
```

Out[10]:



```
In [11]: t=pd.DataFrame(data.groupby(by=data['type'])['isFraud'].sum())
    sns.barplot(data=t,x=t.index, y='isFraud')
```

Out[11]: <AxesSubplot:xlabel='type', ylabel='isFraud'>



```
In [12]: t=pd.DataFrame(data.groupby(by=data['step'])['isFraud'].sum())
    t2=t.sort_values(by='isFraud', ascending=False).head(20)
    t2 = t2.rename_axis('step').reset_index()
    plt.figure(figsize=(20,5))
    sns.barplot(data=t2,x='step', y='isFraud',)
    plt.ylabel('Fraud count by step')
```

Distribution of Transaction Type

```
32.8
```

```
In [14]: # Checking correlation
    correlation = data.corr()
    print(correlation["isFraud"].sort_values(ascending=False))
```

```
0.052494
         amount
         oldbalanceOrg -0.000483
         newbalanceDest -0.001764
         oldbalanceDest -0.005179
         newbalanceOrig -0.006235
                        -0.020222
         step
         isFlaggedFraud
                              NaN
         Name: isFraud, dtype: float64
In [15]:
         #transform the categorical into numeric
         data["type"] = data["type"].map({"CASH_OUT": 1, "PAYMENT": 2,
                                        "CASH_IN": 3, "TRANSFER": 4,
                                        "DEBIT": 5})
         data["isFraud"] = data["isFraud"].map({0: "No Fraud", 1: "Fraud"})
         print(data.head())
                                   nameOrig oldbalanceOrg newbalanceOrig \
           step type
                        amount
         0
                  2 9839.64 C1231006815 170136.0 160296.36
              1
                   2 1864.28 C1666544295
                                                              19384.72
         1
              1
                                                 21249.0
                                                  181.0
                      181.00 C1305486145
         2
              1
                   4
                                                                    0.00
                        181.00 C840083671
                                                    181.0
         3
              1
                    1
                                                                    0.00
                                                 41554.0 29885.86
                    2 11668.14 C2048537720
              nameDest oldbalanceDest newbalanceDest isFraud isFlaggedFraud
         0 M1979787155
                                                 0.0 No Fraud
                                 0.0
                                                                            0
                                                 0.0 No Fraud
         1 M2044282225
                                  0.0
                                                                            0
         2
            C553264065
                                 0.0
                                                 0.0
                                                      Fraud
                                                                            0
             C38997010
                                                 0.0
         3
                             21182.0
                                                        Fraud
                                                                            a
         4 M1230701703
                                                 0.0 No Fraud
                                                                            a
                                 0.0
In [16]: # splitting the data
         from sklearn.model_selection import train_test_split
         x = np.array(data[["type", "amount", "oldbalanceOrg", "newbalanceOrig"]])
         y = np.array(data[["isFraud"]])
In [17]: # training a machine Learning model
         from sklearn.tree import DecisionTreeClassifier
         xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size=0.40, random_state=
         model = DecisionTreeClassifier()
         model.fit(xtrain, ytrain)
         print(model.score(xtest, ytest))
         0.9993338313736873
In [18]: from sklearn.linear_model import LogisticRegression
         lr = LogisticRegression()
         lr.fit(xtrain,ytrain)
         lr.score(xtrain,ytrain)
         C:\Users\Dell\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataCon
         versionWarning:
         A column-vector y was passed when a 1d array was expected. Please change the shape
         of y to (n_samples, ), for example using ravel().
         0.999866765640762
Out[18]:
In [19]: from sklearn.ensemble import RandomForestClassifier
         model = RandomForestClassifier()
         model.fit(xtrain,ytrain)
         model.score(xtrain,ytrain)
```

isFraud

1.000000

C:\Users\Dell\AppData\Local\Temp\ipykernel_14260\2079082207.py:3: DataConversionWa
rning:

A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().

Out[19]: 1.0

In [20]: # prediction
#features = [type, amount, oldbalanceOrg, newbalanceOrig]
features = np.array([[4, 5000.60, 3000.60, 0.0]])
print(model.predict(features))

['No Fraud']

In [21]: #prediction

features=np.array([[1,181.00,181.00,0.0]])
print(model.predict(features))

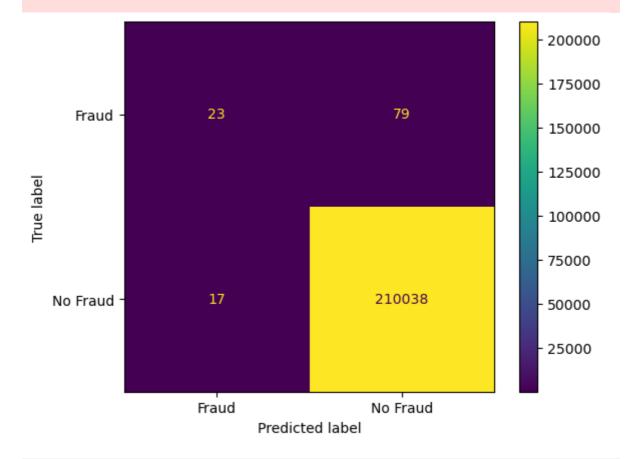
['Fraud']

In [22]: from sklearn.metrics import plot_confusion_matrix

plot_confusion_matrix(model, xtest, ytest)
plt.show()

C:\Users\Dell\anaconda3\lib\site-packages\sklearn\utils\deprecation.py:87: FutureW
arning:

Function plot_confusion_matrix is deprecated; Function `plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one of the class methods: ConfusionMatrixDisplay.from_predictions or ConfusionMatrixDisplay.from_estimator.



In [23]: #import classification_report
 from sklearn.metrics import classification_report

```
In [24]: classifier = DecisionTreeClassifier()
    classifier.fit(xtrain, ytrain)
    ypred = classifier.predict(xtest)
    report = classification_report(ytest, ypred)
    print(report)
```

	precision	recall	f1-score	support
Fraud	0.27	0.25	0.26	102
No Fraud	1.00	1.00	1.00	210055
accuracy			1.00	210157
macro avg	0.63	0.63	0.63	210157
weighted avg	1.00	1.00	1.00	210157

```
In [25]: classifier = RandomForestClassifier()
    classifier.fit(xtrain, ytrain)
    ypred = classifier.predict(xtest)
    report = classification_report(ytest, ypred)
    print(report)
```

C:\Users\Dell\AppData\Local\Temp\ipykernel_14260\803100067.py:2: DataConversionWar
ning:

A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().

	precision	recall	f1-score	support
Fraud	0.59	0.24	0.34	102
No Fraud	1.00	1.00	1.00	210055
accuracy			1.00	210157
macro avg	0.79	0.62	0.67	210157
weighted avg	1.00	1.00	1.00	210157