```
In [1]: import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        import seaborn as sns
In [2]: df = pd.read_csv(r"C:\Users\veera\Downloads\PRANAV B\docs\projects\Transactional
In [3]: print("Shape of data:", df.shape)
        print("\nColumns:", df.columns.tolist())
        print("\nData types:\n", df.dtypes)
        print("\nMissing values:\n", df.isnull().sum().sum())
       Shape of data: (284807, 31)
       Columns: ['Time', 'V1', 'V2', 'V3', 'V4', 'V5', 'V6', 'V7', 'V8', 'V9', 'V10', 'V
       11', 'V12', 'V13', 'V14', 'V15', 'V16', 'V17', 'V18', 'V19', 'V20', 'V21', 'V22',
       'V23', 'V24', 'V25', 'V26', 'V27', 'V28', 'Amount', 'Class']
       Data types:
       Time
                 float64
       V1
                 float64
       V2
                 float64
       V3
                 float64
       V4
                 float64
       V5
                 float64
       V6
                 float64
       V7
                 float64
       ٧8
                 float64
       V9
                 float64
       V10
                 float64
       V11
                 float64
       V12
                 float64
                 float64
       V13
       V14
                 float64
       V15
                 float64
       V16
                 float64
       V17
                 float64
       V18
                 float64
       V19
                 float64
       V20
                 float64
       V21
                 float64
       V22
                 float64
       V23
                 float64
                 float64
       V24
       V25
                 float64
       V26
                 float64
       V27
                 float64
       V28
                 float64
                 float64
       Amount
       Class
                   int64
       dtype: object
       Missing values:
        0
In [4]: | df.head()
```

Out[4]:		Time	V1	V2	V3	V4	V5	V6	V7	
	0	0.0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388	0.239599	0.0
	1	0.0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361	-0.078803	0.0
	2	1.0	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499	0.791461	0.2
	3	1.0	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203	0.237609	0.3
	4	2.0	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921	0.592941	-0.2

5 rows × 31 columns

In [5]: df.describe().T
 #Amount - transaction values
 #Class - 1 = Fraud, 0 = Legit

Out[5]:

	count	mean	std	min	25%	50%
Time	284807.0	9.481386e+04	47488.145955	0.000000	54201.500000	84692.000000
V1	284807.0	1.168375e-15	1.958696	-56.407510	-0.920373	0.018109
V2	284807.0	3.416908e-16	1.651309	-72.715728	-0.598550	0.065486
V3	284807.0	-1.379537e- 15	1.516255	-48.325589	-0.890365	0.179846
V4	284807.0	2.074095e-15	1.415869	-5.683171	-0.848640	-0.019847
V5	284807.0	9.604066e-16	1.380247	-113.743307	-0.691597	-0.054336
V6	284807.0	1.487313e-15	1.332271	-26.160506	-0.768296	-0.274187
V7	284807.0	-5.556467e- 16	1.237094	-43.557242	-0.554076	0.040103
V8	284807.0	1.213481e-16	1.194353	-73.216718	-0.208630	0.022358
V9	284807.0	-2.406331e- 15	1.098632	-13.434066	-0.643098	-0.051429
V10	284807.0	2.239053e-15	1.088850	-24.588262	-0.535426	-0.092917
V11	284807.0	1.673327e-15	1.020713	-4.797473	-0.762494	-0.032757
V12	284807.0	-1.247012e- 15	0.999201	-18.683715	-0.405571	0.140033
V13	284807.0	8.190001e-16	0.995274	-5.791881	-0.648539	-0.013568
V14	284807.0	1.207294e-15	0.958596	-19.214325	-0.425574	0.050601
V15	284807.0	4.887456e-15	0.915316	-4.498945	-0.582884	0.048072
V16	284807.0	1.437716e-15	0.876253	-14.129855	-0.468037	0.066413
V17	284807.0	-3.772171e- 16	0.849337	-25.162799	-0.483748	-0.065676
V18	284807.0	9.564149e-16	0.838176	-9.498746	-0.498850	-0.003636
V19	284807.0	1.039917e-15	0.814041	-7.213527	-0.456299	0.003735
V20	284807.0	6.406204e-16	0.770925	-54.497720	-0.211721	-0.062481
V21	284807.0	1.654067e-16	0.734524	-34.830382	-0.228395	-0.029450
V22	284807.0	-3.568593e- 16	0.725702	-10.933144	-0.542350	0.006782
V23	284807.0	2.578648e-16	0.624460	-44.807735	-0.161846	-0.011193
V24	284807.0	4.473266e-15	0.605647	-2.836627	-0.354586	0.040976
V25	284807.0	5.340915e-16	0.521278	-10.295397	-0.317145	0.016594
V26	284807.0	1.683437e-15	0.482227	-2.604551	-0.326984	-0.052139
V27	284807.0	-3.660091e- 16	0.403632	-22.565679	-0.070840	0.001342

		count	mean	std	min	25%	50%
	V28	284807.0	-1.227390e- 16	0.330083	-15.430084	-0.052960	0.011244
An	nount	284807.0	8.834962e+01	250.120109	0.000000	5.600000	22.000000
	Class	284807.0	1.727486e-03	0.041527	0.000000	0.000000	0.000000

```
In [6]: fraud_count = df['Class'].value_counts()
    print(fraud_count)

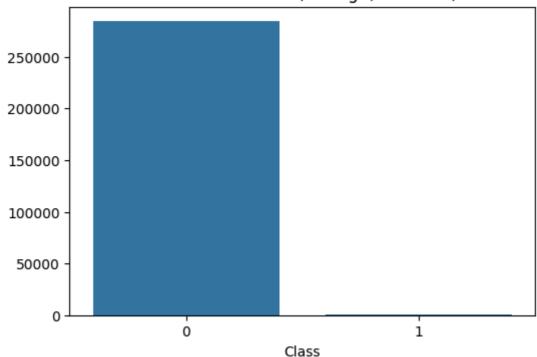
plt.figure(figsize=(6,4))
    sns.barplot(x=fraud_count.index, y=fraud_count.values)
    plt.title("Class Distribution (0=Legit, 1=Fraud)")
    plt.show()

fraud_percentage = (fraud_count[1] / fraud_count.sum()) * 100
    print(f"Fraud Percentage: {fraud_percentage:.4f}%")
```

Class 0 284315 1 492

Name: count, dtype: int64

Class Distribution (0=Legit, 1=Fraud)



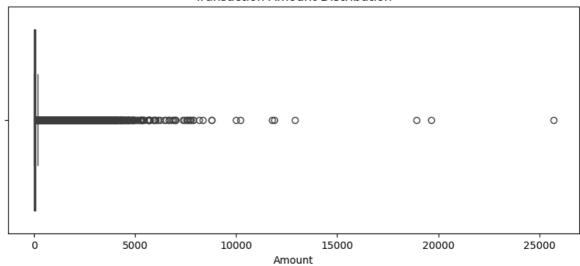
Fraud Percentage: 0.1727%

```
In [7]: # Check for NaNs again
    print(df.isnull().sum().sum())

# Quick outlier visualization for 'Amount'
    plt.figure(figsize=(10,4))
    sns.boxplot(x=df['Amount'])
    plt.title("Transaction Amount Distribution")
    plt.show()
```

0

Transaction Amount Distribution



In [8]: df.to_csv(r"C:\Users\veera\Downloads\PRANAV B\docs\projects\Transactional Fraud
print("Clean data saved.")

Clean data saved.

In []: !jupyter nbconvert --to html "week1_data_exploration.ipynb" --output-dir="report