

```
import kagglehub
path = kagglehub.dataset_download("cristobaltudela/credit-card-transaction-1

n outdated `kagglehub` version (installed: 0.3.13), please consider upgrading
e.com/api/v1/datasets/download/cristobaltudela/credit-card-transaction-legiti
0<00:00, 82.5MB/s]Extracting files...
```

```
from google.colab import files
uploaded = files.upload()
```

```
Choose Files fraud_detection.csv
fraud_detection.csv(text/csv) - 324710 bytes, last modified: 1/30/2026 - 100% done
Saving fraud_detection.csv to fraud_detection.csv
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

▼ New section

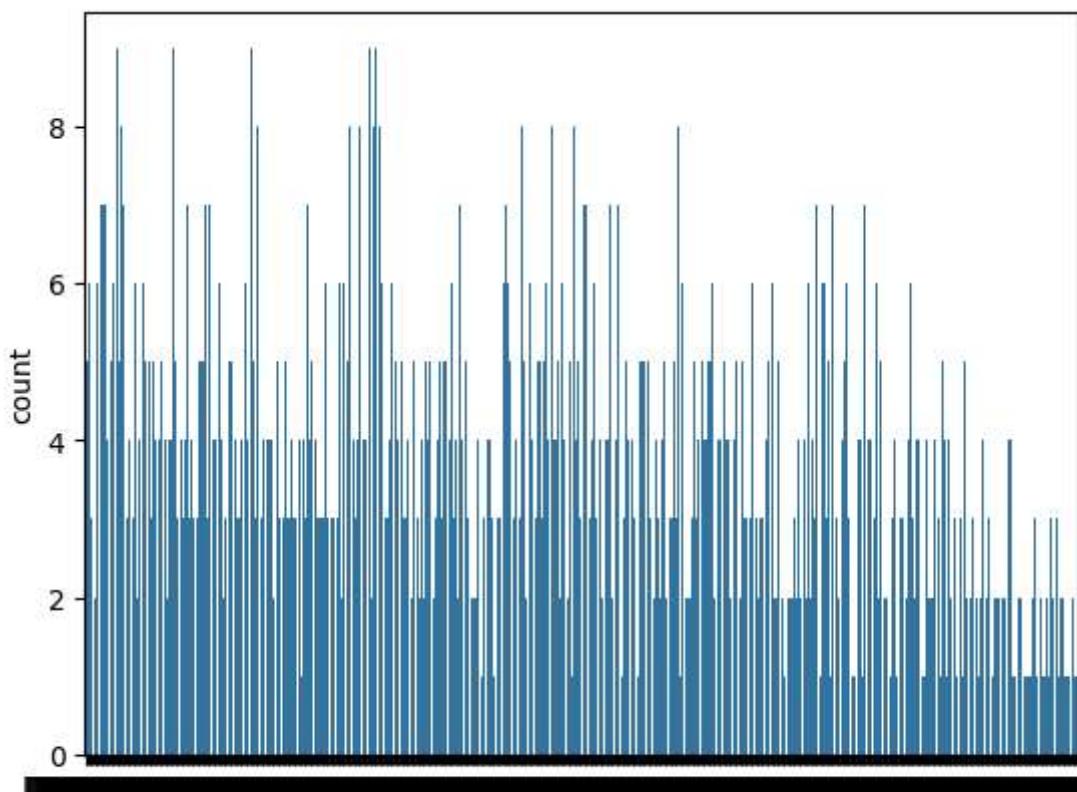
```
#Step 2. Load data
df = pd.read_csv("/content/fraud_detection.csv")
df.head()
#Step 3. Basic checks
df.shape
print(df.columns)
#df['TransactionID'].value_counts() # This line caused the error. After insp

Index(['TransactionID', 'AccountID', 'TransactionAmount', 'TransactionDate',
       'TransactionType', 'Location', 'DeviceID', 'IP Address', 'MerchantID',
       'Channel', 'CustomerAge', 'CustomerOccupation', 'TransactionDuration',
       'LoginAttempts', 'AccountBalance', 'PreviousTransactionDate'],
      dtype='object')
```

```
#Step 4. Data understanding
```

```
#Check class imbalance:
```

```
sns.countplot(x='DeviceID', data=df) # 'Class' column not found. Please repl
plt.show()
```



```
print(df.columns)
```

```
Index(['TransactionID', 'AccountID', 'TransactionAmount', 'TransactionDate',
       'TransactionType', 'Location', 'DeviceID', 'IP Address', 'MerchantID',
       'Channel', 'CustomerAge', 'CustomerOccupation', 'TransactionDuration',
       'LoginAttempts', 'AccountBalance', 'PreviousTransactionDate'],
      dtype='object')
```

```
# Step-by-step EDA and Outlier Handling on the uploaded dataset
```

```
# 1. Basic inspection
```

```
shape = df.shape
```

```
shape, head
```

```
((2512, 16),
 TransactionID AccountID TransactionAmount TransactionDate \
 0 TX000001 AC00128 14.09 4/11/2023 16:29
 1 TX000002 AC00455 376.24 6/27/2023 16:44
 2 TX000003 AC00019 126.29 7/10/2023 18:16
 3 TX000004 AC00070 184.50 5/5/2023 16:32
 4 TX000005 AC00411 13.45 10/16/2023 17:51
```

	TransactionType	Location	DeviceID	IP Address	MerchantID	Channel
0	Debit	San Diego	D000380	162.198.218.92	M015	ATM
1	Debit	Houston	D000051	13.149.61.4	M052	ATM
2	Debit	Mesa	D000235	215.97.143.157	M009	Online

```

3          Debit    Raleigh  D000187  200.13.225.150      M002  Online
4          Credit   Atlanta  D000308   65.164.3.100      M091  Online

   CustomerAge CustomerOccupation TransactionDuration  LoginAttempts \
0            70             Doctor                  81           1
1            68             Doctor                 141           1
2            19            Student                  56           1
3            26            Student                  25           1
4            26            Student                 198           1

   AccountBalance PreviousTransactionDate
0        5112.21           11/4/2024 8:08
1       13758.91           11/4/2024 8:09
2        1122.35           11/4/2024 8:07
3       8569.06           11/4/2024 8:09
4       7429.40           11/4/2024 8:06 )

```

```

info = df.info()
head = df.head()

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2512 entries, 0 to 2511
Data columns (total 16 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   TransactionID    2512 non-null   object 
 1   AccountID        2512 non-null   object 
 2   TransactionAmount 2512 non-null   float64
 3   TransactionDate   2512 non-null   object 
 4   TransactionType   2512 non-null   object 
 5   Location          2512 non-null   object 
 6   DeviceID          2512 non-null   object 
 7   IP Address         2512 non-null   object 
 8   MerchantID        2512 non-null   object 
 9   Channel           2512 non-null   object 
 10  CustomerAge       2512 non-null   int64  
 11  CustomerOccupation 2512 non-null   object 
 12  TransactionDuration 2512 non-null   int64  
 13  LoginAttempts     2512 non-null   int64  
 14  AccountBalance    2512 non-null   float64
 15  PreviousTransactionDate 2512 non-null   object 
dtypes: float64(2), int64(3), object(11)
memory usage: 314.1+ KB

```

```

# 2. Descriptive statistics
print("\nDescriptive Statistics:")
print(df.describe(include="all"))

```

```

Descriptive Statistics:
   TransactionID  AccountID  TransactionAmount  TransactionDate \
count          2512        2512      2512.000000          2512
unique         2512        495                  NaN          2405
top          TX002496      AC00460                  NaN  11/20/2023 16:29

```

freq	1	12		NaN		3
mean		NaN	NaN	297.593778		NaN
std		NaN	NaN	291.946243		NaN
min		NaN	NaN	0.260000		NaN
25%		NaN	NaN	81.885000		NaN
50%		NaN	NaN	211.140000		NaN
75%		NaN	NaN	414.527500		NaN
max		NaN	NaN	1919.110000		NaN
	TransactionType	Location	DeviceID	IP Address	MerchantID	\
count	2512	2512	2512		2512	2512
unique	2	43	681		592	100
top	Debit	Fort Worth	D000548	200.136.146.93		M026
freq	1944	70	9		13	45
mean		NaN	NaN		NaN	NaN
std		NaN	NaN		NaN	NaN
min		NaN	NaN		NaN	NaN
25%		NaN	NaN		NaN	NaN
50%		NaN	NaN		NaN	NaN
75%		NaN	NaN		NaN	NaN
max		NaN	NaN		NaN	NaN
	Channel	CustomerAge	CustomerOccupation	TransactionDuration		\
count	2512	2512.000000		2512		2512.000000
unique	3	Nan		4		Nan
top	Branch	Nan		Student		Nan
freq	868	Nan		657		Nan
mean		44.673965		Nan		119.643312
std		17.792198		Nan		69.963757
min		18.000000		Nan		10.000000
25%		27.000000		Nan		63.000000
50%		45.000000		Nan		112.500000
75%		59.000000		Nan		161.000000
max		80.000000		Nan		300.000000
	LoginAttempts	AccountBalance	PreviousTransactionDate			
count	2512.000000	2512.000000			2512	
unique		Nan	Nan			7
top		Nan	Nan	11/4/2024 8:07		
freq		Nan	Nan			435
mean	1.124602	5114.302966				Nan
std	0.602662	3900.942499				Nan
min	1.000000	101.250000				Nan
25%	1.000000	1504.370000				Nan
50%	1.000000	4735.510000				Nan
75%	1.000000	7678.820000				Nan
max	5.000000	14977.990000				Nan

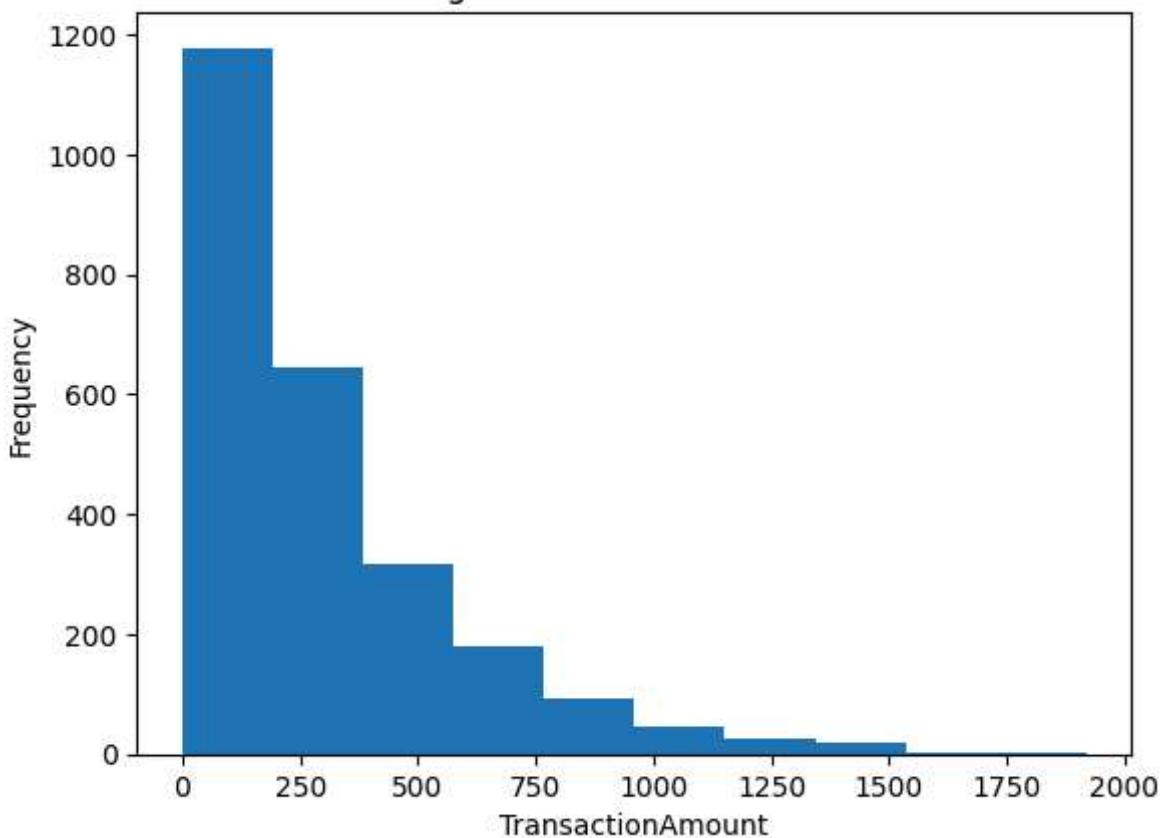
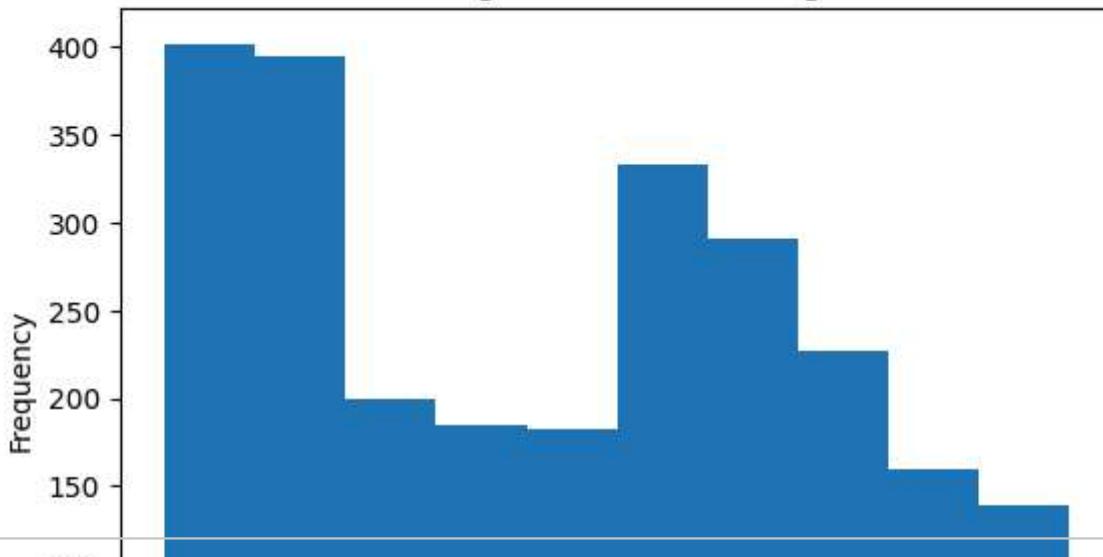
```
# 3. Missing value percentage
missing_pct = df.isnull().mean() * 100
print("\nMissing Value Percentage (%):")
print(missing_pct)
```

Missing Value Percentage (%):

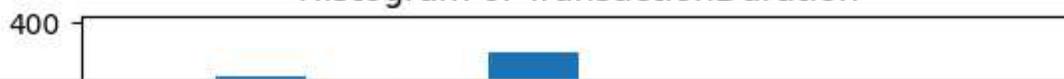
```
TransactionID      0.0
AccountId         0.0
TransactionAmount 0.0
TransactionDate   0.0
TransactionType   0.0
Location          0.0
DeviceID          0.0
IP Address        0.0
MerchantID        0.0
Channel           0.0
CustomerAge       0.0
CustomerOccupation 0.0
TransactionDuration 0.0
LoginAttempts     0.0
AccountBalance    0.0
PreviousTransactionDate 0.0
dtype: float64
```

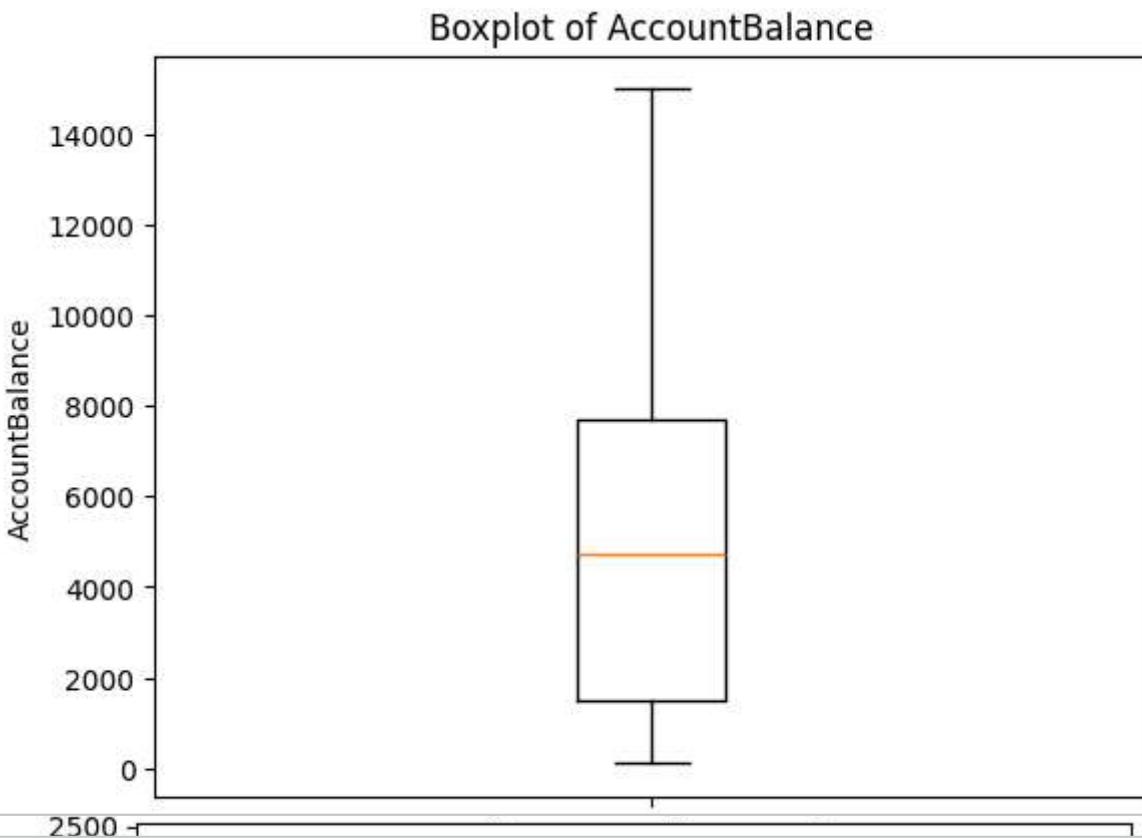
```
# 4. Plot distributions (numeric columns only)
numeric_cols = df.select_dtypes(include=np.number).columns

for col in numeric_cols:
    plt.figure()
    plt.hist(df[col].dropna())
    plt.title(f"Histogram of {col}")
    plt.xlabel(col)
    plt.ylabel("Frequency")
    plt.show()
```


Histogram of TransactionAmount**Histogram of CustomerAge**

```
plt.figure()
plt.boxplot(df[col].dropna(), vert=True)
plt.title(f"Boxplot of {col}")
plt.ylabel(col)
plt.show()
```

Histogram of TransactionDuration

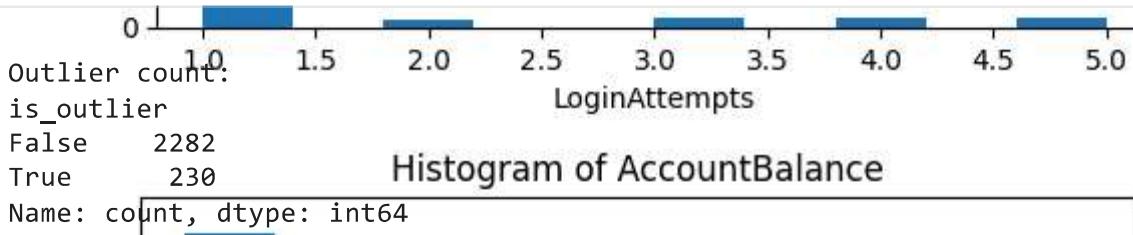


```
# 5. Detect outliers using IQR
outlier_flags = pd.DataFrame(index=df.index)

for col in numeric_cols:
    Q1 = df[col].quantile(0.25)
    Q3 = df[col].quantile(0.75)
    IQR = Q3 - Q1
    lower = Q1 - 1.5 * IQR
    upper = Q3 + 1.5 * IQR
    outlier_flags[col + "_outlier"] = ((df[col] < lower) | (df[col] > upper))
```

```
# 6. Create overall outlier flag column
df["is_outlier"] = outlier_flags.any(axis=1)

print("\nOutlier count:")
print(df["is_outlier"].value_counts())
```



```
# 7. Handle outliers: capping (winsorization) to preserve data size
df_clean = df.copy()
```

```

for col in numeric_cols:
    Q1 = df_clean[col].quantile(0.25)
    Q3 = df_clean[col].quantile(0.75)
    IQR = Q3 - Q1
    lower = Q1 - 1.5 * IQR
    upper = Q3 + 1.5 * IQR
    df_clean[col] = np.where(df_clean[col] < lower, lower, df_clean[col])
    df_clean[col] = np.where(df_clean[col] > upper, upper, df_clean[col])

```

```

# 8. Correlation matrix
corr_matrix = df_clean[numeric_cols].corr()
print("\nCorrelation Matrix:")
print(corr_matrix)

```

Correlation Matrix:

	TransactionAmount	CustomerAge	TransactionDuration	\
TransactionAmount	1.000000	-0.021949	0.002129	
CustomerAge	-0.021949	1.000000	-0.017936	
TransactionDuration	0.002129	-0.017936	1.000000	
LoginAttempts		NaN	NaN	NaN
AccountBalance	-0.021299	0.319942	0.005577	
	LoginAttempts	AccountBalance		
TransactionAmount	NaN	-0.021299		
CustomerAge	NaN	0.319942		
TransactionDuration	NaN	0.005577		
LoginAttempts	NaN		NaN	
AccountBalance	NaN	1.000000		

```

# Top correlations (absolute value)
corr_pairs = (
    corr_matrix.abs()
    .unstack()
    .sort_values(ascending=False)
)

top_corr = corr_pairs[corr_pairs < 1].head(5)
print("\nTop Correlations:")
print(top_corr)

```

Top Correlations:

CustomerAge	AccountBalance	0.319942
AccountBalance	CustomerAge	0.319942
CustomerAge	TransactionAmount	0.021949
TransactionAmount	CustomerAge	0.021949
	AccountBalance	0.021299

dtype: float64