

# DATA Visualization ASSIGNMENT

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A red ink signature, possibly reading 'Simi', with a decorative flourish underneath.

## Banking Transaction Dataset

### \* Sample Dataset

Customer-ID	Name	Age	Acc. type	T.ID	Date	Amo	Balance
C001	Rahul Sharma	28	Savings	T001	25-09-20	2500	15000
C002	Rupq Singh	35	Current	T002	25-9-2	10000	50000
C003	Aarav Mehta	30	Savings	T003	15-9-3	2500	25000
C004	Neha Patel	40	Current	T004	25-9-4	20000	100000
C005	Riya Gupta	22	Savings	T005	25-9-5	1500	12000
C006	Karan Verma	45	Savings	T006	25-9-6	5000	42000
C007	Meena Iyer	32	Current	T007	25-9-7	6000	35000

a1) Show the following details using the given dataset

a) Highest Transaction amount, lowest transaction amount

We need to find -

Maximum value of Amo → Highest

Minimum value of Amo → lowest

```
import pandas as pd
```

```
df = pd.read_csv('banking-dataset.csv')
```

```
highest = df['Amo'].max()
```

```
lowest = df['Amo'].min()
```

```
Print ("Highest Transaction Amount:", highest)
```

```
Print ("Lowest Transaction Amount:", lowest)
```

(b) Count of customers who made more than 10 transaction in a month

We group the dataset by Customer ID and month, then the number of Transactions. Finally Filter those having more than 10 transactions.

```
df['month'].pd.to_datetime(df['Date']).dt.month  
transaction_count = df.groupby(['Customer ID', 'month'])  
    .size().reset(index=True, name='Transaction-  
count')
```

active\_customers = transaction\_count[transaction\_count['Transaction-count'] > 10]

```
Print('Customers with more than 10 transactions in a month:')  
Print(active_customers)
```

(c) Display All customers Names who have Savings Account type.

- We need to filter all records where Acc-type = 'Savings' and show unique customer Name

```
Savings_count = df[df['Acc-type'] == 'Savings']['Name'].unique  
Print('Customer having Savings Account:')
```

for name in Savings\_customers:  
 Print(name)

OUTPUT:-

Highest Amount: 20000

Lowest Amount: 1500

Customers having Savings:

Rahul Sharma

Annu Mehta

Riya Gupta

Ratan Verma

Customers with more than

10 transactions: Non in this sample

Q2) Construct a Histogram for Univariate Analysis, Scatter plot for Bivariate (Transaction Vs Balance), and pairplot for Multivariate (Transaction, Balance, Age).

a) Histogram for Univariate Analysis (Transaction Amount)

- Purpose is to observe the distribution of Transaction amount, detect skewness or outliers.

```
plt.figure(figsize=(7,5))
```

```
plt.hist(df['Amount'], bins=5, colour='skyblue', edgecolor='black')
```

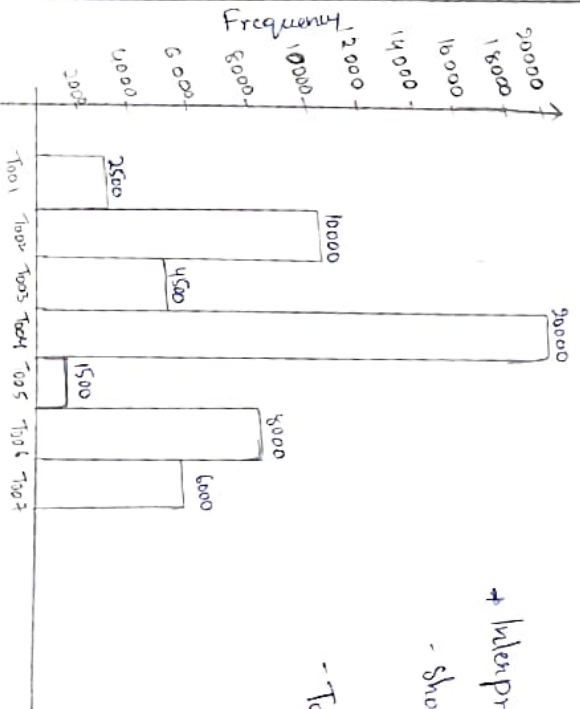
```
plt.title('Histogram of Transaction Amounts')
```

```
plt.xlabel('Transaction Amount')
```

```
plt.ylabel('Frequency')
```

```
plt.grid(axis='y', linestyle='--', alpha=0.7)
```

```
plt.show()
```



\* Interpretation

- Shows how amount spread  
- Taller bars, more Amount

Transactions - ID

## (b) Scatterplot for Bivariate Analysis (Transaction Vs Bal.)

- Purpose : to visualize the relationship between Amo vs B.  
: Identify positive / Negative correlation

- Code :-

```
plt.figure(figsize=(7,5))
```

```
sns.scatterplot(x='Amount', y='Balance',
```

```
data=df, s=100, colour='teal')
```

```
plt.title("Scatterplot: Transaction Vs Balance")
```

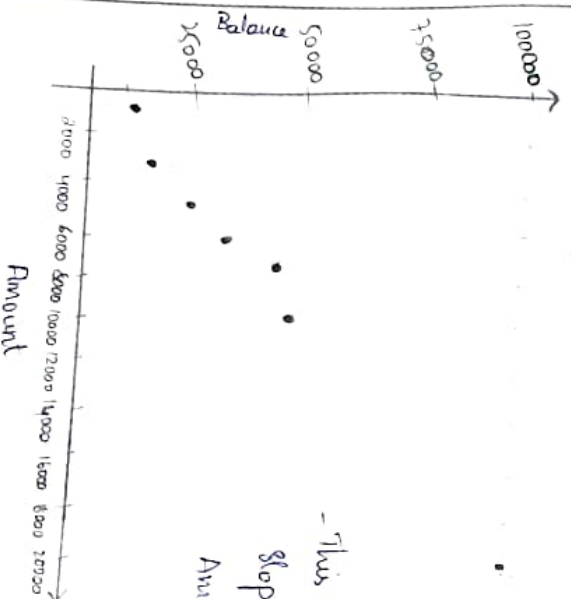
```
plt.xlabel("Transaction Amount")
```

```
plt.ylabel("Balance Amount")
```

```
plt.grid(True, linestyle='--', alpha=0.6)
```

```
plt.show()
```

Scatter plot: Transaction Vs Balance



- This has a positive trend  
Slope upward, higher  
Amount, higher balance.

(c) Pairplot for Multi variable Analysis (Transaction, Balance, Age)

- Purpose:- Analyse Multiple relationships simultaneously among numeric variables. Helps detect patterns or clusters

- code:-

sns.pairplot(df[['Age', 'Balance', 'Age17', 'diag', 'kind', 'kde'],

corner=True)

plt.figure(figsize=(10, 5))

plt.show()

plt.figure(figsize=(10, 5))

o3) Utilize WordCloud package to generate a cloud of Transaction remarks / comments and visualize frequency of transaction

types (Deposit / Withdrawal).

\* Sample dataset

Trans - type :-	Deposit	Withdrawal	Deposit	Withdrawal	Deposit	Withdrawal
Trans - Remarks :-	Salary Credited	Bill Payment	Money transfer to friend	Shopping Expense	Mobile recharge	Interest Credited
						Online purchase

- WordCloud:- Visualize most frequent words from Transaction type (Deposit / Withdrawal)

Code:-

text = ""

wordcloud = WordCloud(width=800, height=400, background-

color='white').generate(text)

plt.figure(figsize=(10, 5))

plt.imshow(wordcloud, interpolation='bilinear')



```
plt.axis("off")
```

```
plt.title('Word cloud of Transaction Remarks', fontsize=6)
```

```
plt.show()
```

- Countplot/Bar chart :- to get the Frequency of Transaction

```
plt.figure(figsize=(6,4))
```

```
sns.countplot(x='Transaction_type', data=df, palette='set2')
```

```
plt.title('Frequency of Transaction Types (Deposit- Vs Withdrawal)')
```

```
plt.xlabel('Transaction Type')
```

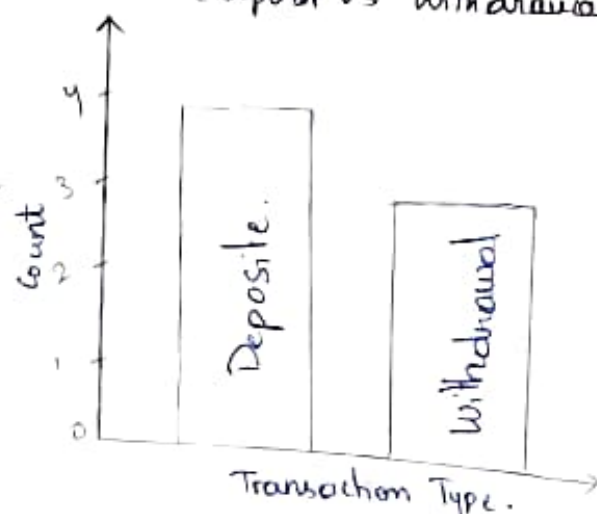
```
plt.ylabel('count')
```

```
plt.show()
```

Word cloud

Payment  
Money  
Bill  
online Shopping  
Interest recharge  
Mobile  
Salary  
recharge  
transfer  
friend  
Purchase

Frequency of Trans. Types  
(Deposit Vs Withdrawal)



Q4) Build a Geospatial heatmap showing branch locations and transaction density across cities

- Geospatial Visualization allows us to represent data with a geographic components

- Heatmaps Visually encode magnitude or density using Color intensity.

## \* Extended Sample Dataset

Branch-ID	Name	City	Lat	Long	Total-Trans	Total amount
Boo1	Central Branch	Mumbai	19.076	72.877	120	250000
Boo2	South Branch	Chennai	13.082	80.270	90	160000
Boo3	East Branch	Kolkata	22.572	88.363	110	200000
Boo4	North Branch	Delhi	28.613	77.209	150	300000
Boo5	West Branch	Bengaluru	12.971	77.594	100	220000

- code:-

```
m = folium.Map(location = [22.5, 78.9], zoom_start = 5, tiles = 'cartodb positron')
```

```
for i, row in df.iterrows():
```

```
    folium.Marker(location=[row['latitude'], row['longitude']],
                  popup=f'{row["Branch-Name"]} ({row["City"]})  

                  3) In Transaction: {row["Total-Transactions"]}  

                  In Amount: ₹ {row["Total-Amount"]}',
                  icon=folium.Icon(color='blue', icon='bank', prefix='fa')).add_to(m)
```

```
heat_data = [[row['latitude'], row['latitude'], row['Total-Amount']]  

              for index, row in df.iterrows()]  

HeatMap(heat_data, radius=25, blur=15, max_zoom=6), add_to(m)
```

```
m.save("bank-branch-heatmap.html")
```



Q5) Analyze and visualize customer transaction trends over time using line Graph (Monthly deposits vs withdrawals)

- Time Series Analysis: Helps identify trends, patterns and seasonality in transaction data.
- Line Graph: Ideal for visualizing changes in numerical data over time
- Matplotlib / Seaborn: Used for creating clear, continuous plots of transactions vs. time.

Month:	1	2	3	4	5	6	7	8	9	10	11
Deposits:	120000	135000	140000	160000	155000	170000	180000	175000	165000	160000	150000
Withdrawals	90000	95000	100000	110000	120000	125000	130000	135000	128000	120000	115000

- Code:-

```
plt.figure(figsize=(10,6))
```

```
sns.lineplot(x='Month', y='Deposits', data=df, marker='o',)
```

```
sns.lineplot(x='Month', y='Withdrawals', data=df, marker='s',  
label='Withdrawals', color='red')
```

```
plt.title('Monthly Deposits Vs Withdrawals Trend', fontsize=14)
```

```
plt.xlabel('Month')
```

```
plt.ylabel('Transaction Amount (₹)')
```

```
plt.legend()
```

```
plt.grid(True, linestyle='--', alpha=0.6)
```

```
plt.tight_layout()
```

```
plt.show()
```