

```
In [75]: import pandas as pd
```

```
In [76]: df=pd.read_csv("C:/Users/NITISH/OneDrive/Documents/Desktop/Bank Data/TOP BANKS.csv")
```

```
In [77]: print(df.head())
```

	Gender	City	Age group	Occupation	BANK1 \
0	Male	Mumbai	31-40	Own Business	HDFC Bank
1	Female	Bangalore	41-50	Salaried & Private	ICICI Bank
2	Male	Ahmedabad	41-50	Own Business	Kotak Mahindra Bank
3	Male	Bangalore	26-30	Salaried & Private	HDFC Bank
4	Male	Ahmedabad	51-60	Own Business	Kotak Mahindra Bank

	BANK2	BANK3	Kotak Mahindra Bank	HDFC Bank	ICICI Bank \
0	ICICI Bank	Yes Bank	Kotak Mahindra Bank	-	-
1	HDFC Bank	Kotak Mahindra Bank	Kotak Mahindra Bank	-	-
2	HDFC Bank	Axis Bank	-	-	-
3	ICICI Bank	Kotak Mahindra Bank	-	HDFC Bank	-
4	HDFC Bank	State Bank Of India	Kotak Mahindra Bank	-	-

	Axis Bank	IndusInd Bank	RBL Bank	Other 1	Other 2 \
0	-	-	-	-	-
1	-	-	-	-	-
2	-	IndusInd Bank	-	South Indian Bank	-
3	-	-	-	Bank Of India	-
4	-	-	-	-	-

	Salary Acc	Primary Acc
0	Yes Bank	Yes Bank
1	Bank Of Baroda	HDFC Bank
2	Axis Bank	Kotak Mahindra Bank
3	None of the above	ICICI Bank
4	Kotak Mahindra Bank	Kotak Mahindra Bank

```
In [78]: Age_group = ['18-25', '26-30', '31-40', '41-50', '51-60']
```

```
In [79]: df['Age group'] = df['Age group'].replace({
    '31-40': '31-40',
    '41-50': '41-50',
    '51-60': '51-60',
    '26-35': '26-30',
    '18-25': '18-25'
})
```

```
In [80]: gender_distribution = df['Gender'].value_counts()
```

```
In [81]: city_distribution = df['City'].value_counts()
age_group_distribution = df['Age group'].value_counts()
```

```
In [82]: all_banks = df.iloc[:, 4:15] # Selecting columns corresponding to banks
bank_counts = all_banks.apply(pd.Series.value_counts).sum(axis=1)
most_preferred_banks = bank_counts.sort_values(ascending=False).head()
```

```
In [83]: occupation_banks = df.groupby('Occupation').agg({
    'BANK1': lambda x: x.value_counts().index[0],
    'BANK2': lambda x: x.value_counts().index[0],
    'BANK3': lambda x: x.value_counts().index[0]
})
```

```
In [84]: primary_salary_accounts = df['Primary Acc'].value_counts()
```

```
In [85]: print("Distribution of respondents by gender:\n", gender_distribution)
print("\nDistribution of respondents by city:\n", city_distribution)
print("\nDistribution of respondents by age group:\n", age_group_distribution)
print("\nMost preferred banks among respondents:\n", most_preferred_banks)
print("\nOccupation-wise preferred banks:\n", occupation_banks)
print("\nDistribution of primary salary accounts among banks:\n", primary_salary_ac
```

## Distribution of respondents by gender:

Gender

Male 255

Female 200

Name: count, dtype: int64

## Distribution of respondents by city:

City

Mumbai 77

Delhi 61

Kolkata 57

Ahmedabad 54

Lucknow 51

Bangalore 46

Chennai 42

Indore 27

Kochi 24

Bhubaneswar 16

Name: count, dtype: int64

## Distribution of respondents by age group:

Age group

18-25 134

31-40 121

26-30 116

41-50 65

51-60 19

Name: count, dtype: int64

## Most preferred banks among respondents:

- 2895.0

HDFC Bank 447.0

Kotak Mahindra Bank 370.0

ICICI Bank 334.0

State Bank Of India 229.0

dtype: float64

## Occupation-wise preferred banks:

	BANK1	BANK2 \
Occupation		
Housewife	HDFC Bank	State Bank Of India
Own Business	HDFC Bank	HDFC Bank
Salaried - Government	Kotak Mahindra Bank	Bank Of Baroda
Salaried [?] Private	HDFC Bank	HDFC Bank
Student	ICICI Bank	State Bank Of India

## BANK3

Occupation	
Housewife	ICICI Bank
Own Business	State Bank Of India
Salaried - Government	ICICI Bank
Salaried [?] Private	ICICI Bank
Student	HDFC Bank

## Distribution of primary salary accounts among banks:

Primary Acc

HDFC Bank 121

Kotak Mahindra Bank 120

ICICI Bank 90

Axis Bank 48

State Bank Of India 17

IndusInd Bank 16

Punjab National Bank 5

RBL Bank 5

Union Bank of India	4
-	4
Canara Bank	4
Yes Bank	4
Bank Of Baroda	3
Bank Of India	1
Karnataka Bank	1
Indian Bank	1
South Indian Bank	1
Equitas Small Finance Bank	1
IDBI Bank	1
DBS Bank	1
Paytm Payments Bank	1
IDFC Bank	1
HSBC India	1
Kerala Bank	1
Tamilnad Mercantile Bank	1
Federal Bank	1
Bank Of Maharashtra	1

Name: count, dtype: int64

```
In [86]: categorical_vars = ['Gender', 'City', 'Age group', 'Occupation', 'Primary Acc', 'Sal']
freq_distribution = {}
for col in categorical_vars:
    freq_distribution[col] = df[col].value_counts()
```

```
In [87]: numerical_vars = df.select_dtypes(include=['int64', 'float64']).columns
correlation_matrix = df[numerical_vars].corr()
```

```
In [88]: from scipy.stats import chi2_contingency
```

```
In [89]: summary_stats = df.describe()

observed = pd.crosstab(df['Gender'], df['Age group'])
chi2, p, _, _ = chi2_contingency(observed)
if p < 0.05:
    hypothesis_result = "Reject null hypothesis: There is a significant difference."
else:
    hypothesis_result = "Fail to reject null hypothesis: There is no significant di"
```

```
In [90]: print("Summary statistics:\n", summary_stats)
print("\nFrequency distribution for categorical variables:\n")
for col, freq_dist in freq_distribution.items():
    print(col + ":\n", freq_dist)
print("\nCorrelation matrix:\n", correlation_matrix)
print("\nHypothesis testing result:\n", hypothesis_result)
```

## Summary statistics:

	Gender	City	Age group	Occupation	BANK1	BANK2 \
count	455	455	455	455	455	453
unique	2	10	5	5	23	32
top	Male	Mumbai	18-25	Salaried & Private	HDFC Bank	HDFC Bank
freq	255	77	134	300	139	77

	BANK3 Kotak Mahindra Bank	HDFC Bank	ICICI Bank \
count	449	455	455
unique	35	2	2
top	State Bank Of India	-	-
freq	62	277	279

	Axis Bank	IndusInd Bank	RBL Bank	Other 1	Other 2	Salary Acc \
count	455	455	455	455	455	455
unique	2	2	2	24	7	22
top	-	-	-	-	-	None of the above
freq	375	424	446	320	443	134

	Primary Acc
count	455
unique	27
top	HDFC Bank
freq	121

## Frequency distribution for categorical variables:

## Gender:

Gender

Male 255

Female 200

Name: count, dtype: int64

## City:

City

Mumbai 77

Delhi 61

Kolkata 57

Ahmedabad 54

Lucknow 51

Bangalore 46

Chennai 42

Indore 27

Kochi 24

Bhubaneswar 16

Name: count, dtype: int64

## Age group:

Age group

18-25 134

31-40 121

26-30 116

41-50 65

51-60 19

Name: count, dtype: int64

## Occupation:

Occupation

Salaried &amp; Private 300

Own Business 73

Housewife 67

Student 14

Salaried - Government 1

Name: count, dtype: int64

## Primary Acc:

Primary Acc

HDFC Bank 121

Kotak Mahindra Bank	120
ICICI Bank	90
Axis Bank	48
State Bank Of India	17
IndusInd Bank	16
Punjab National Bank	5
RBL Bank	5
Union Bank of India	4
-	4
Canara Bank	4
Yes Bank	4
Bank Of Baroda	3
Bank Of India	1
Karnataka Bank	1
Indian Bank	1
South Indian Bank	1
Equitas Small Finance Bank	1
IDBI Bank	1
DBS Bank	1
Paytm Payments Bank	1
IDFC Bank	1
HSBC India	1
Kerala Bank	1
Tamilnad Mercantile Bank	1
Federal Bank	1
Bank Of Maharashtra	1

Name: count, dtype: int64

Salary Acc:

Salary Acc	
None of the above	134
HDFC Bank	91
Kotak Mahindra Bank	73
ICICI Bank	71
Axis Bank	38
IndusInd Bank	13
State Bank Of India	8
Bank Of Baroda	6
Punjab National Bank	4
Canara Bank	2
Union Bank of India	2
Yes Bank	2
Bank Of India	2
HSBC India	1
Bank Of Maharashtra	1
Federal Bank	1
DBS Bank	1
IDFC Bank	1
Equitas Small Finance Bank	1
IDBI Bank	1
South Indian Bank	1
Indian Bank	1

Name: count, dtype: int64

Correlation matrix:

Empty DataFrame

Columns: []

Index: []

Hypothesis testing result:

Fail to reject null hypothesis: There is no significant difference.

```
In [91]: primary_salary_bank1 = pd.crosstab(df['Primary Acc'], df['Salary Acc'], margins=True)
primary_salary_bank2 = pd.crosstab(df['Primary Acc'], df['Salary Acc'], margins=True)
primary_salary_bank3 = pd.crosstab(df['Primary Acc'], df['Salary Acc'], margins=True)

In [92]: print("Cross-tabulation of Primary Acc and Salary Acc with Bank1:\n", primary_salary_bank1)
print("\nCross-tabulation of Primary Acc and Salary Acc with Bank2:\n", primary_salary_bank2)
print("\nCross-tabulation of Primary Acc and Salary Acc with Bank3:\n", primary_salary_bank3)
```

## Cross-tabulation of Primary Acc and Salary Acc with Bank1:

Salary Acc	Axis Bank	Bank Of Baroda	Bank Of India \
Primary Acc			
-	0	0	0
Axis Bank	29	0	0
Bank Of Baroda	0	2	0
Bank Of India	0	0	1
Bank Of Maharashtra	0	0	0
Canara Bank	0	0	0
DBS Bank	0	0	0
Equitas Small Finance Bank	0	0	0
Federal Bank	0	0	0
HDFC Bank	3	2	1
HSBC India	0	0	0
ICICI Bank	0	0	0
IDBI Bank	1	0	0
IDFC Bank	0	0	0
Indian Bank	0	0	0
IndusInd Bank	0	0	0
Karnataka Bank	0	0	0
Kerala Bank	1	0	0
Kotak Mahindra Bank	1	2	0
Paytm Payments Bank	0	0	0
Punjab National Bank	0	0	0
RBL Bank	0	0	0
South Indian Bank	0	0	0
State Bank Of India	3	0	0
Tamilnad Mercantile Bank	0	0	0
Union Bank of India	0	0	0
Yes Bank	0	0	0
All	38	6	2

Salary Acc	Bank Of Maharashtra	Canara Bank	DBS Bank \
Primary Acc			
-	0	0	0
Axis Bank	0	0	0
Bank Of Baroda	0	0	0
Bank Of India	0	0	0
Bank Of Maharashtra	1	0	0
Canara Bank	0	2	0
DBS Bank	0	0	0
Equitas Small Finance Bank	0	0	0
Federal Bank	0	0	0
HDFC Bank	0	0	0
HSBC India	0	0	0
ICICI Bank	0	0	0
IDBI Bank	0	0	0
IDFC Bank	0	0	0
Indian Bank	0	0	0
IndusInd Bank	0	0	0
Karnataka Bank	0	0	0
Kerala Bank	0	0	0
Kotak Mahindra Bank	0	0	1
Paytm Payments Bank	0	0	0
Punjab National Bank	0	0	0
RBL Bank	0	0	0
South Indian Bank	0	0	0
State Bank Of India	0	0	0
Tamilnad Mercantile Bank	0	0	0
Union Bank of India	0	0	0
Yes Bank	0	0	0
All	1	2	1

Salary Acc	Equitas Small Finance Bank	Federal Bank \
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Primary Acc					
-			0		0
Axis Bank			0		0
Bank Of Baroda			0		0
Bank Of India			0		0
Bank Of Maharashtra			0		0
Canara Bank			0		0
DBS Bank			0		0
Equitas Small Finance Bank			1		0
Federal Bank			0		1
HDFC Bank			0		0
HSBC India			0		0
ICICI Bank			0		0
IDBI Bank			0		0
IDFC Bank			0		0
Indian Bank			0		0
IndusInd Bank			0		0
Karnataka Bank			0		0
Kerala Bank			0		0
Kotak Mahindra Bank			0		0
Paytm Payments Bank			0		0
Punjab National Bank			0		0
RBL Bank			0		0
South Indian Bank			0		0
State Bank Of India			0		0
Tamilnad Mercantile Bank			0		0
Union Bank of India			0		0
Yes Bank			0		0
All			1		1
Salary Acc	HDFC Bank	HSBC India	...	Indian Bank	\
Primary Acc			...		
-	1	0	...		0
Axis Bank	0	0	...		0
Bank Of Baroda	0	0	...		0
Bank Of India	0	0	...		0
Bank Of Maharashtra	0	0	...		0
Canara Bank	1	0	...		0
DBS Bank	0	0	...		0
Equitas Small Finance Bank	0	0	...		0
Federal Bank	0	0	...		0
HDFC Bank	78	0	...		0
HSBC India	0	1	...		0
ICICI Bank	3	0	...		0
IDBI Bank	0	0	...		0
IDFC Bank	1	0	...		0
Indian Bank	0	0	...		0
IndusInd Bank	0	0	...		1
Karnataka Bank	0	0	...		0
Kerala Bank	0	0	...		0
Kotak Mahindra Bank	1	0	...		0
Paytm Payments Bank	0	0	...		0
Punjab National Bank	0	0	...		0
RBL Bank	2	0	...		0
South Indian Bank	0	0	...		0
State Bank Of India	3	0	...		0
Tamilnad Mercantile Bank	0	0	...		0
Union Bank of India	1	0	...		0
Yes Bank	0	0	...		0
All	91	1	...		1
Salary Acc	IndusInd Bank	Kotak Mahindra Bank	\		
Primary Acc					
-	0				0

Axis Bank	0	0
Bank Of Baroda	0	0
Bank Of India	0	0
Bank Of Maharashtra	0	0
Canara Bank	0	0
DBS Bank	0	0
Equitas Small Finance Bank	0	0
Federal Bank	0	0
HDFC Bank	1	2
HSBC India	0	0
ICICI Bank	1	0
IDBI Bank	0	0
IDFC Bank	0	0
Indian Bank	0	0
IndusInd Bank	11	0
Karnataka Bank	0	0
Kerala Bank	0	0
Kotak Mahindra Bank	0	68
Paytm Payments Bank	0	0
Punjab National Bank	0	2
RBL Bank	0	0
South Indian Bank	0	0
State Bank Of India	0	0
Tamilnad Mercantile Bank	0	0
Union Bank of India	0	0
Yes Bank	0	1
All	13	73

Salary Acc                      None of the above    Punjab National Bank    \

Primary Acc

-	2	0
Axis Bank	18	0
Bank Of Baroda	1	0
Bank Of India	0	0
Bank Of Maharashtra	0	0
Canara Bank	1	0
DBS Bank	1	0
Equitas Small Finance Bank	0	0
Federal Bank	0	0
HDFC Bank	31	0
HSBC India	0	0
ICICI Bank	19	0
IDBI Bank	0	0
IDFC Bank	0	0
Indian Bank	1	0
IndusInd Bank	3	0
Karnataka Bank	1	0
Kerala Bank	0	0
Kotak Mahindra Bank	43	1
Paytm Payments Bank	1	0
Punjab National Bank	0	3
RBL Bank	3	0
South Indian Bank	0	0
State Bank Of India	5	0
Tamilnad Mercantile Bank	1	0
Union Bank of India	2	0
Yes Bank	1	0
All	134	4

Salary Acc                      South Indian Bank    State Bank Of India    \

Primary Acc

-	0	0
Axis Bank	0	0
Bank Of Baroda	0	0

Bank Of India	0	0
Bank Of Maharashtra	0	0
Canara Bank	0	0
DBS Bank	0	0
Equitas Small Finance Bank	0	0
Federal Bank	0	0
HDFC Bank	0	1
HSBC India	0	0
ICICI Bank	0	1
IDBI Bank	0	0
IDFC Bank	0	0
Indian Bank	0	0
IndusInd Bank	0	0
Karnataka Bank	0	0
Kerala Bank	0	0
Kotak Mahindra Bank	0	1
Paytm Payments Bank	0	0
Punjab National Bank	0	0
RBL Bank	0	0
South Indian Bank	1	0
State Bank Of India	0	5
Tamilnad Mercantile Bank	0	0
Union Bank of India	0	0
Yes Bank	0	0
All	1	8

Salary Acc	Union Bank of India	Yes Bank	All
Primary Acc			
-	0	0	4
Axis Bank	0	0	48
Bank Of Baroda	0	0	3
Bank Of India	0	0	1
Bank Of Maharashtra	0	0	1
Canara Bank	0	0	4
DBS Bank	0	0	1
Equitas Small Finance Bank	0	0	1
Federal Bank	0	0	1
HDFC Bank	0	0	121
HSBC India	0	0	1
ICICI Bank	1	0	90
IDBI Bank	0	0	1
IDFC Bank	0	0	1
Indian Bank	0	0	1
IndusInd Bank	0	0	16
Karnataka Bank	0	0	1
Kerala Bank	0	0	1
Kotak Mahindra Bank	0	0	120
Paytm Payments Bank	0	0	1
Punjab National Bank	0	0	5
RBL Bank	0	0	5
South Indian Bank	0	0	1
State Bank Of India	0	0	17
Tamilnad Mercantile Bank	0	0	1
Union Bank of India	1	0	4
Yes Bank	0	2	4
All	2	2	455

[28 rows x 23 columns]

Cross-tabulation of Primary Acc and Salary Acc with Bank2:

Salary Acc	Axis Bank	Bank Of Baroda	Bank Of India \
Primary Acc			
-	0	0	0
Axis Bank	29	0	0

Bank Of Baroda	0	2	0
Bank Of India	0	0	1
Bank Of Maharashtra	0	0	0
Canara Bank	0	0	0
DBS Bank	0	0	0
Equitas Small Finance Bank	0	0	0
Federal Bank	0	0	0
HDFC Bank	3	2	1
HSBC India	0	0	0
ICICI Bank	0	0	0
IDBI Bank	1	0	0
IDFC Bank	0	0	0
Indian Bank	0	0	0
IndusInd Bank	0	0	0
Karnataka Bank	0	0	0
Kerala Bank	1	0	0
Kotak Mahindra Bank	1	2	0
Paytm Payments Bank	0	0	0
Punjab National Bank	0	0	0
RBL Bank	0	0	0
South Indian Bank	0	0	0
State Bank Of India	3	0	0
Tamilnad Mercantile Bank	0	0	0
Union Bank of India	0	0	0
Yes Bank	0	0	0
All	38	6	2

Salary Acc	Bank Of Maharashtra	Canara Bank	DBS Bank	\
Primary Acc				
-	0	0	0	
Axis Bank	0	0	0	
Bank Of Baroda	0	0	0	
Bank Of India	0	0	0	
Bank Of Maharashtra	1	0	0	
Canara Bank	0	2	0	
DBS Bank	0	0	0	
Equitas Small Finance Bank	0	0	0	
Federal Bank	0	0	0	
HDFC Bank	0	0	0	
HSBC India	0	0	0	
ICICI Bank	0	0	0	
IDBI Bank	0	0	0	
IDFC Bank	0	0	0	
Indian Bank	0	0	0	
IndusInd Bank	0	0	0	
Karnataka Bank	0	0	0	
Kerala Bank	0	0	0	
Kotak Mahindra Bank	0	0	1	
Paytm Payments Bank	0	0	0	
Punjab National Bank	0	0	0	
RBL Bank	0	0	0	
South Indian Bank	0	0	0	
State Bank Of India	0	0	0	
Tamilnad Mercantile Bank	0	0	0	
Union Bank of India	0	0	0	
Yes Bank	0	0	0	
All	1	2	1	

Salary Acc	Equitas Small Finance Bank	Federal Bank	\
Primary Acc			
-	0	0	
Axis Bank	0	0	
Bank Of Baroda	0	0	
Bank Of India	0	0	

Bank Of Maharashtra	0	0
Canara Bank	0	0
DBS Bank	0	0
Equitas Small Finance Bank	1	0
Federal Bank	0	1
HDFC Bank	0	0
HSBC India	0	0
ICICI Bank	0	0
IDBI Bank	0	0
IDFC Bank	0	0
Indian Bank	0	0
IndusInd Bank	0	0
Karnataka Bank	0	0
Kerala Bank	0	0
Kotak Mahindra Bank	0	0
Paytm Payments Bank	0	0
Punjab National Bank	0	0
RBL Bank	0	0
South Indian Bank	0	0
State Bank Of India	0	0
Tamilnad Mercantile Bank	0	0
Union Bank of India	0	0
Yes Bank	0	0
All	1	1

Salary Acc	HDFC Bank	HSBC India	...	Indian Bank	\
Primary Acc			...		
-	1	0	...	0	
Axis Bank	0	0	...	0	
Bank Of Baroda	0	0	...	0	
Bank Of India	0	0	...	0	
Bank Of Maharashtra	0	0	...	0	
Canara Bank	1	0	...	0	
DBS Bank	0	0	...	0	
Equitas Small Finance Bank	0	0	...	0	
Federal Bank	0	0	...	0	
HDFC Bank	78	0	...	0	
HSBC India	0	1	...	0	
ICICI Bank	3	0	...	0	
IDBI Bank	0	0	...	0	
IDFC Bank	1	0	...	0	
Indian Bank	0	0	...	0	
IndusInd Bank	0	0	...	1	
Karnataka Bank	0	0	...	0	
Kerala Bank	0	0	...	0	
Kotak Mahindra Bank	1	0	...	0	
Paytm Payments Bank	0	0	...	0	
Punjab National Bank	0	0	...	0	
RBL Bank	2	0	...	0	
South Indian Bank	0	0	...	0	
State Bank Of India	3	0	...	0	
Tamilnad Mercantile Bank	0	0	...	0	
Union Bank of India	1	0	...	0	
Yes Bank	0	0	...	0	
All	91	1	...	1	

Salary Acc	IndusInd Bank	Kotak Mahindra Bank	\
Primary Acc			
-	0	0	
Axis Bank	0	0	
Bank Of Baroda	0	0	
Bank Of India	0	0	
Bank Of Maharashtra	0	0	
Canara Bank	0	0	

DBS Bank	0	0
Equitas Small Finance Bank	0	0
Federal Bank	0	0
HDFC Bank	1	2
HSBC India	0	0
ICICI Bank	1	0
IDBI Bank	0	0
IDFC Bank	0	0
Indian Bank	0	0
IndusInd Bank	11	0
Karnataka Bank	0	0
Kerala Bank	0	0
Kotak Mahindra Bank	0	68
Paytm Payments Bank	0	0
Punjab National Bank	0	2
RBL Bank	0	0
South Indian Bank	0	0
State Bank Of India	0	0
Tamilnad Mercantile Bank	0	0
Union Bank of India	0	0
Yes Bank	0	1
All	13	73

Salary Acc	None of the above	Punjab National Bank \
Primary Acc		
-	2	0
Axis Bank	18	0
Bank Of Baroda	1	0
Bank Of India	0	0
Bank Of Maharashtra	0	0
Canara Bank	1	0
DBS Bank	1	0
Equitas Small Finance Bank	0	0
Federal Bank	0	0
HDFC Bank	31	0
HSBC India	0	0
ICICI Bank	19	0
IDBI Bank	0	0
IDFC Bank	0	0
Indian Bank	1	0
IndusInd Bank	3	0
Karnataka Bank	1	0
Kerala Bank	0	0
Kotak Mahindra Bank	43	1
Paytm Payments Bank	1	0
Punjab National Bank	0	3
RBL Bank	3	0
South Indian Bank	0	0
State Bank Of India	5	0
Tamilnad Mercantile Bank	1	0
Union Bank of India	2	0
Yes Bank	1	0
All	134	4

Salary Acc	South Indian Bank	State Bank Of India \
Primary Acc		
-	0	0
Axis Bank	0	0
Bank Of Baroda	0	0
Bank Of India	0	0
Bank Of Maharashtra	0	0
Canara Bank	0	0
DBS Bank	0	0
Equitas Small Finance Bank	0	0

Federal Bank	0	0
HDFC Bank	0	1
HSBC India	0	0
ICICI Bank	0	1
IDBI Bank	0	0
IDFC Bank	0	0
Indian Bank	0	0
IndusInd Bank	0	0
Karnataka Bank	0	0
Kerala Bank	0	0
Kotak Mahindra Bank	0	1
Paytm Payments Bank	0	0
Punjab National Bank	0	0
RBL Bank	0	0
South Indian Bank	1	0
State Bank Of India	0	5
Tamilnad Mercantile Bank	0	0
Union Bank of India	0	0
Yes Bank	0	0
All	1	8

Salary Acc	Union Bank of India	Yes Bank	All
Primary Acc			
-	0	0	4
Axis Bank	0	0	48
Bank Of Baroda	0	0	3
Bank Of India	0	0	1
Bank Of Maharashtra	0	0	1
Canara Bank	0	0	4
DBS Bank	0	0	1
Equitas Small Finance Bank	0	0	1
Federal Bank	0	0	1
HDFC Bank	0	0	121
HSBC India	0	0	1
ICICI Bank	1	0	90
IDBI Bank	0	0	1
IDFC Bank	0	0	1
Indian Bank	0	0	1
IndusInd Bank	0	0	16
Karnataka Bank	0	0	1
Kerala Bank	0	0	1
Kotak Mahindra Bank	0	0	120
Paytm Payments Bank	0	0	1
Punjab National Bank	0	0	5
RBL Bank	0	0	5
South Indian Bank	0	0	1
State Bank Of India	0	0	17
Tamilnad Mercantile Bank	0	0	1
Union Bank of India	1	0	4
Yes Bank	0	2	4
All	2	2	455

[28 rows x 23 columns]

Cross-tabulation of Primary Acc and Salary Acc with Bank3:

Salary Acc	Axis Bank	Bank Of Baroda	Bank Of India \
Primary Acc			
-	0	0	0
Axis Bank	29	0	0
Bank Of Baroda	0	2	0
Bank Of India	0	0	1
Bank Of Maharashtra	0	0	0
Canara Bank	0	0	0
DBS Bank	0	0	0

Equitas Small Finance Bank	0	0	0
Federal Bank	0	0	0
HDFC Bank	3	2	1
HSBC India	0	0	0
ICICI Bank	0	0	0
IDBI Bank	1	0	0
IDFC Bank	0	0	0
Indian Bank	0	0	0
IndusInd Bank	0	0	0
Karnataka Bank	0	0	0
Kerala Bank	1	0	0
Kotak Mahindra Bank	1	2	0
Paytm Payments Bank	0	0	0
Punjab National Bank	0	0	0
RBL Bank	0	0	0
South Indian Bank	0	0	0
State Bank Of India	3	0	0
Tamilnad Mercantile Bank	0	0	0
Union Bank of India	0	0	0
Yes Bank	0	0	0
All	38	6	2

Salary Acc	Bank Of Maharashtra	Canara Bank	DBS Bank	\
Primary Acc				
-	0	0	0	
Axis Bank	0	0	0	
Bank Of Baroda	0	0	0	
Bank Of India	0	0	0	
Bank Of Maharashtra	1	0	0	
Canara Bank	0	2	0	
DBS Bank	0	0	0	
Equitas Small Finance Bank	0	0	0	
Federal Bank	0	0	0	
HDFC Bank	0	0	0	
HSBC India	0	0	0	
ICICI Bank	0	0	0	
IDBI Bank	0	0	0	
IDFC Bank	0	0	0	
Indian Bank	0	0	0	
IndusInd Bank	0	0	0	
Karnataka Bank	0	0	0	
Kerala Bank	0	0	0	
Kotak Mahindra Bank	0	0	1	
Paytm Payments Bank	0	0	0	
Punjab National Bank	0	0	0	
RBL Bank	0	0	0	
South Indian Bank	0	0	0	
State Bank Of India	0	0	0	
Tamilnad Mercantile Bank	0	0	0	
Union Bank of India	0	0	0	
Yes Bank	0	0	0	
All	1	2	1	

Salary Acc	Equitas Small Finance Bank	Federal Bank	\
Primary Acc			
-	0	0	
Axis Bank	0	0	
Bank Of Baroda	0	0	
Bank Of India	0	0	
Bank Of Maharashtra	0	0	
Canara Bank	0	0	
DBS Bank	0	0	
Equitas Small Finance Bank	1	0	
Federal Bank	0	1	



HDFC Bank	0	0
HSBC India	0	0
ICICI Bank	0	0
IDBI Bank	0	0
IDFC Bank	0	0
Indian Bank	0	0
IndusInd Bank	0	0
Karnataka Bank	0	0
Kerala Bank	0	0
Kotak Mahindra Bank	0	0
Paytm Payments Bank	0	0
Punjab National Bank	0	0
RBL Bank	0	0
South Indian Bank	0	0
State Bank Of India	0	0
Tamilnad Mercantile Bank	0	0
Union Bank of India	0	0
Yes Bank	0	0
All	1	1

Salary Acc	HDFC Bank	HSBC India	...	Indian Bank	\
Primary Acc			...		
-	1	0	...	0	
Axis Bank	0	0	...	0	
Bank Of Baroda	0	0	...	0	
Bank Of India	0	0	...	0	
Bank Of Maharashtra	0	0	...	0	
Canara Bank	1	0	...	0	
DBS Bank	0	0	...	0	
Equitas Small Finance Bank	0	0	...	0	
Federal Bank	0	0	...	0	
HDFC Bank	78	0	...	0	
HSBC India	0	1	...	0	
ICICI Bank	3	0	...	0	
IDBI Bank	0	0	...	0	
IDFC Bank	1	0	...	0	
Indian Bank	0	0	...	0	
IndusInd Bank	0	0	...	1	
Karnataka Bank	0	0	...	0	
Kerala Bank	0	0	...	0	
Kotak Mahindra Bank	1	0	...	0	
Paytm Payments Bank	0	0	...	0	
Punjab National Bank	0	0	...	0	
RBL Bank	2	0	...	0	
South Indian Bank	0	0	...	0	
State Bank Of India	3	0	...	0	
Tamilnad Mercantile Bank	0	0	...	0	
Union Bank of India	1	0	...	0	
Yes Bank	0	0	...	0	
All	91	1	...	1	

Salary Acc	IndusInd Bank	Kotak Mahindra Bank	\
Primary Acc			
-	0	0	
Axis Bank	0	0	
Bank Of Baroda	0	0	
Bank Of India	0	0	
Bank Of Maharashtra	0	0	
Canara Bank	0	0	
DBS Bank	0	0	
Equitas Small Finance Bank	0	0	
Federal Bank	0	0	
HDFC Bank	1	2	
HSBC India	0	0	

ICICI Bank	1	0
IDBI Bank	0	0
IDFC Bank	0	0
Indian Bank	0	0
IndusInd Bank	11	0
Karnataka Bank	0	0
Kerala Bank	0	0
Kotak Mahindra Bank	0	68
Paytm Payments Bank	0	0
Punjab National Bank	0	2
RBL Bank	0	0
South Indian Bank	0	0
State Bank Of India	0	0
Tamilnad Mercantile Bank	0	0
Union Bank of India	0	0
Yes Bank	0	1
All	13	73

Salary Acc                      None of the above    Punjab National Bank    \

Primary Acc		
-	2	0
Axis Bank	18	0
Bank Of Baroda	1	0
Bank Of India	0	0
Bank Of Maharashtra	0	0
Canara Bank	1	0
DBS Bank	1	0
Equitas Small Finance Bank	0	0
Federal Bank	0	0
HDFC Bank	31	0
HSBC India	0	0
ICICI Bank	19	0
IDBI Bank	0	0
IDFC Bank	0	0
Indian Bank	1	0
IndusInd Bank	3	0
Karnataka Bank	1	0
Kerala Bank	0	0
Kotak Mahindra Bank	43	1
Paytm Payments Bank	1	0
Punjab National Bank	0	3
RBL Bank	3	0
South Indian Bank	0	0
State Bank Of India	5	0
Tamilnad Mercantile Bank	1	0
Union Bank of India	2	0
Yes Bank	1	0
All	134	4

Salary Acc                      South Indian Bank    State Bank Of India    \

Primary Acc		
-	0	0
Axis Bank	0	0
Bank Of Baroda	0	0
Bank Of India	0	0
Bank Of Maharashtra	0	0
Canara Bank	0	0
DBS Bank	0	0
Equitas Small Finance Bank	0	0
Federal Bank	0	0
HDFC Bank	0	1
HSBC India	0	0
ICICI Bank	0	1
IDBI Bank	0	0

IDFC Bank	0	0
Indian Bank	0	0
IndusInd Bank	0	0
Karnataka Bank	0	0
Kerala Bank	0	0
Kotak Mahindra Bank	0	1
Paytm Payments Bank	0	0
Punjab National Bank	0	0
RBL Bank	0	0
South Indian Bank	1	0
State Bank Of India	0	5
Tamilnad Mercantile Bank	0	0
Union Bank of India	0	0
Yes Bank	0	0
All	1	8

Salary Acc	Union Bank of India	Yes Bank	All
Primary Acc			
-	0	0	4
Axis Bank	0	0	48
Bank Of Baroda	0	0	3
Bank Of India	0	0	1
Bank Of Maharashtra	0	0	1
Canara Bank	0	0	4
DBS Bank	0	0	1
Equitas Small Finance Bank	0	0	1
Federal Bank	0	0	1
HDFC Bank	0	0	121
HSBC India	0	0	1
ICICI Bank	1	0	90
IDBI Bank	0	0	1
IDFC Bank	0	0	1
Indian Bank	0	0	1
IndusInd Bank	0	0	16
Karnataka Bank	0	0	1
Kerala Bank	0	0	1
Kotak Mahindra Bank	0	0	120
Paytm Payments Bank	0	0	1
Punjab National Bank	0	0	5
RBL Bank	0	0	5
South Indian Bank	0	0	1
State Bank Of India	0	0	17
Tamilnad Mercantile Bank	0	0	1
Union Bank of India	1	0	4
Yes Bank	0	2	4
All	2	2	455

[28 rows x 23 columns]

```
In [93]: import pandas as pd
import numpy as np
import statsmodels.api as sm
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error
```

```
In [94]: print("Summary statistics:")
print(df.describe())
```

Summary statistics:

	Gender	City	Age group	Occupation	BANK1	BANK2	\
count	455	455	455	455	455	453	
unique	2	10	5	5	23	32	
top	Male	Mumbai	18-25	Salaried & Private	HDFC Bank	HDFC Bank	
freq	255	77	134	300	139	77	

	BANK3	Kotak Mahindra Bank	HDFC Bank	ICICI Bank	\
count	449	455	455	455	
unique	35	2	2	2	
top	State Bank Of India	-	-	-	
freq	62	277	279	331	

	Axis Bank	IndusInd Bank	RBL Bank	Other 1	Other 2	Salary Acc	\
count	455	455	455	455	455	455	
unique	2	2	2	24	7	22	
top	-	-	-	-	-	None of the above	
freq	375	424	446	320	443	134	

	Primary Acc
count	455
unique	27
top	HDFC Bank
freq	121

```
In [95]: print("Missing values:")
print(df.isnull().sum())
```

Missing values:

Gender	0
City	0
Age group	0
Occupation	0
BANK1	0
BANK2	2
BANK3	6
Kotak Mahindra Bank	0
HDFC Bank	0
ICICI Bank	0
Axis Bank	0
IndusInd Bank	0
RBL Bank	0
Other 1	0
Other 2	0
Salary Acc	0
Primary Acc	0

dtype: int64

```
In [96]: numeric_df = df.select_dtypes(include=np.number)
correlation_matrix = numeric_df.corr()
```

```
In [97]: non_numeric_columns = df.select_dtypes(exclude=np.number).columns
print("Non-numeric columns:", non_numeric_columns)
```

```
Non-numeric columns: Index(['Gender', 'City', 'Age group', 'Occupation', 'BANK1',
'BANK2', 'BANK3',
'Kotak Mahindra Bank', 'HDFC Bank', 'ICICI Bank', 'Axis Bank',
'IndusInd Bank', 'RBL Bank', 'Other 1', 'Other 2', 'Salary Acc',
'Primary Acc'],
dtype='object')
```

```
In [98]: encoded_df = pd.get_dummies(df)
```

```
In [99]: corr_matrix = encoded_df.corr()
```

```
In [100]: print("Correlation matrix:")  
print(corr_matrix)
```

Correlation matrix:

	Gender_Female	Gender_Male	\
Gender_Female	1.000000	-1.000000	
Gender_Male	-1.000000	1.000000	
City_Ahmedabad	0.003611	-0.003611	
City_Bangalore	-0.003228	0.003228	
City_Bhubaneswar	-0.000793	0.000793	
...	...	...	
Primary Acc_South Indian Bank	-0.041564	0.041564	
Primary Acc_State Bank Of India	0.012316	-0.012316	
Primary Acc_Tamilnad Mercantile Bank	-0.041564	0.041564	
Primary Acc_Union Bank of India	-0.035968	0.035968	
Primary Acc_Yes Bank	0.058904	-0.058904	
	City_Ahmedabad	City_Bangalore	\
Gender_Female	0.003611	-0.003228	
Gender_Male	-0.003611	0.003228	
City_Ahmedabad	1.000000	-0.123067	
City_Bangalore	-0.123067	1.000000	
City_Bhubaneswar	-0.070057	-0.064024	
...	...	...	
Primary Acc_South Indian Bank	-0.017223	-0.015739	
Primary Acc_State Bank Of India	0.071036	-0.027628	
Primary Acc_Tamilnad Mercantile Bank	-0.017223	0.139944	
Primary Acc_Union Bank of India	-0.034559	0.046517	
Primary Acc_Yes Bank	0.038239	-0.031583	
	City_Bhubaneswar	City_Chennai	\
Gender_Female	-0.000793	0.008237	
Gender_Male	0.000793	-0.008237	
City_Ahmedabad	-0.070057	-0.117024	
City_Bangalore	-0.064024	-0.106947	
City_Bhubaneswar	1.000000	-0.060880	
...	...	...	
Primary Acc_South Indian Bank	-0.008960	-0.014967	
Primary Acc_State Bank Of India	-0.037611	-0.062826	
Primary Acc_Tamilnad Mercantile Bank	-0.008960	-0.014967	
Primary Acc_Union Bank of India	-0.017979	0.132643	
Primary Acc_Yes Bank	-0.017979	-0.030032	
	City_Delhi	City_Indore	City_Kochi \
Gender_Female	0.015424	0.002472	-0.090124
Gender_Male	-0.015424	-0.002472	0.090124
City_Ahmedabad	-0.144391	-0.092169	-0.086595
City_Bangalore	-0.131957	-0.084232	-0.079138
City_Bhubaneswar	-0.075118	-0.047950	-0.045050
...	...	...	...
Primary Acc_South Indian Bank	-0.018467	-0.011788	0.198886
Primary Acc_State Bank Of India	-0.043506	0.048620	0.057200
Primary Acc_Tamilnad Mercantile Bank	-0.018467	-0.011788	-0.011075
Primary Acc_Union Bank of India	-0.037056	-0.023654	-0.022223
Primary Acc_Yes Bank	-0.037056	-0.023654	-0.022223
	City_Kolkata	...	\
Gender_Female	0.119655	...	
Gender_Male	-0.119655	...	
City_Ahmedabad	-0.138874	...	
City_Bangalore	-0.126915	...	
City_Bhubaneswar	-0.072248	...	
...	...	...	
Primary Acc_South Indian Bank	-0.017761	...	
Primary Acc_State Bank Of India	0.030469	...	
Primary Acc_Tamilnad Mercantile Bank	-0.017761	...	
Primary Acc_Union Bank of India	-0.035640	...	

Primary Acc\_Yes Bank

-0.035640 ...

Primary Acc\_Kerala Bank \

Gender_Female	-0.041564
Gender_Male	0.041564
City_Ahmedabad	-0.017223
City_Bangalore	-0.015739
City_Bhubaneswar	-0.008960
...	...
Primary Acc_South Indian Bank	-0.002203
Primary Acc_State Bank Of India	-0.009246
Primary Acc_Tamilnad Mercantile Bank	-0.002203
Primary Acc_Union Bank of India	-0.004420
Primary Acc_Yes Bank	-0.004420

Primary Acc\_Kotak Mahindra Bank \

Gender_Female	-0.007509
Gender_Male	0.007509
City_Ahmedabad	-0.019150
City_Bangalore	-0.068360
City_Bhubaneswar	0.075281
...	...
Primary Acc_South Indian Bank	-0.028089
Primary Acc_State Bank Of India	-0.117911
Primary Acc_Tamilnad Mercantile Bank	-0.028089
Primary Acc_Union Bank of India	-0.056365
Primary Acc_Yes Bank	-0.056365

Primary Acc\_Paytm Payments Bank \

Gender_Female	0.052994
Gender_Male	-0.052994
City_Ahmedabad	-0.017223
City_Bangalore	-0.015739
City_Bhubaneswar	-0.008960
...	...
Primary Acc_South Indian Bank	-0.002203
Primary Acc_State Bank Of India	-0.009246
Primary Acc_Tamilnad Mercantile Bank	-0.002203
Primary Acc_Union Bank of India	-0.004420
Primary Acc_Yes Bank	-0.004420

Primary Acc\_Punjab National Bank \

Gender_Female	-0.050877
Gender_Male	0.050877
City_Ahmedabad	-0.038682
City_Bangalore	0.034582
City_Bhubaneswar	-0.020124
...	...
Primary Acc_South Indian Bank	-0.004947
Primary Acc_State Bank Of India	-0.020767
Primary Acc_Tamilnad Mercantile Bank	-0.004947
Primary Acc_Union Bank of India	-0.009927
Primary Acc_Yes Bank	-0.009927

Primary Acc\_RBL Bank \

Gender_Female	0.034073
Gender_Male	-0.034073
City_Ahmedabad	0.026504
City_Bangalore	0.034582
City_Bhubaneswar	-0.020124
...	...
Primary Acc_South Indian Bank	-0.004947
Primary Acc_State Bank Of India	-0.020767
Primary Acc_Tamilnad Mercantile Bank	-0.004947

Primary Acc_Union Bank of India	-0.009927
Primary Acc_Yes Bank	-0.009927
Primary Acc_South Indian Bank \	
Gender_Female	-0.041564
Gender_Male	0.041564
City_Ahmedabad	-0.017223
City_Bangalore	-0.015739
City_Bhubaneswar	-0.008960
...	...
Primary Acc_South Indian Bank	1.000000
Primary Acc_State Bank Of India	-0.009246
Primary Acc_Tamilnad Mercantile Bank	-0.002203
Primary Acc_Union Bank of India	-0.004420
Primary Acc_Yes Bank	-0.004420
Primary Acc_State Bank Of India \	
Gender_Female	0.012316
Gender_Male	-0.012316
City_Ahmedabad	0.071036
City_Bangalore	-0.027628
City_Bhubaneswar	-0.037611
...	...
Primary Acc_South Indian Bank	-0.009246
Primary Acc_State Bank Of India	1.000000
Primary Acc_Tamilnad Mercantile Bank	-0.009246
Primary Acc_Union Bank of India	-0.018554
Primary Acc_Yes Bank	-0.018554
Primary Acc_Tamilnad Mercantile Bank \	
Gender_Female	-0.041564
Gender_Male	0.041564
City_Ahmedabad	-0.017223
City_Bangalore	0.139944
City_Bhubaneswar	-0.008960
...	...
Primary Acc_South Indian Bank	-0.002203
Primary Acc_State Bank Of India	-0.009246
Primary Acc_Tamilnad Mercantile Bank	1.000000
Primary Acc_Union Bank of India	-0.004420
Primary Acc_Yes Bank	-0.004420
Primary Acc_Union Bank of India \	
Gender_Female	-0.035968
Gender_Male	0.035968
City_Ahmedabad	-0.034559
City_Bangalore	0.046517
City_Bhubaneswar	-0.017979
...	...
Primary Acc_South Indian Bank	-0.004420
Primary Acc_State Bank Of India	-0.018554
Primary Acc_Tamilnad Mercantile Bank	-0.004420
Primary Acc_Union Bank of India	1.000000
Primary Acc_Yes Bank	-0.008869
Primary Acc_Yes Bank	
Gender_Female	0.058904
Gender_Male	-0.058904
City_Ahmedabad	0.038239
City_Bangalore	-0.031583
City_Bhubaneswar	-0.017979
...	...
Primary Acc_South Indian Bank	-0.004420
Primary Acc_State Bank Of India	-0.018554



Primary Acc_Tamilnad Mercantile Bank	-0.004420
Primary Acc_Union Bank of India	-0.008869
Primary Acc_Yes Bank	1.000000

[204 rows x 204 columns]

```
In [101... from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import classification_report
from sklearn.preprocessing import OneHotEncoder
```

```
In [102... X = df.drop(columns=['Primary Acc', 'Salary Acc'])
y = df['Primary Acc']
```

```
In [103... X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
In [104... encoder = OneHotEncoder()
```

```
In [105... encoder = OneHotEncoder(drop='first', sparse=False)
```

```
In [106... combined_data = pd.concat([X_train, X_test])
```

```
In [107... combined_encoded = encoder.fit_transform(combined_data)
```

```
C:\Users\NITISH\anaconda3\Lib\site-packages\sklearn\preprocessing\_encoders.py:97:
FutureWarning: `sparse` was renamed to `sparse_output` in version 1.2 and will
be removed in 1.4. `sparse_output` is ignored unless you leave `sparse` to its default
value.
warnings.warn(
```

```
In [108... X_train_encoded = combined_encoded[:len(X_train)]
X_test_encoded = combined_encoded[len(X_train):]
```

```
In [109... from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import classification_report
```

```
In [110... from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import OneHotEncoder
```

```
In [111... categorical_cols = ['Gender', 'City', 'Age group', 'Occupation', 'BANK1', 'BANK2',
encoder = ColumnTransformer(transformers=[('onehot', OneHotEncoder()), categorical_cols])
```

```
In [112... X_encoded = encoder.fit_transform(df)
```

```
In [113... X_train, X_test, y_train, y_test = train_test_split(X_encoded, y, test_size=0.2, random_state=42)
```

```
In [114... rf_classifier = RandomForestClassifier(n_estimators=100, random_state=42)
rf_classifier.fit(X_train, y_train)
```

```
Out[114]:
RandomForestClassifier
RandomForestClassifier(random_state=42)
```

```
In [115... y_pred = rf_classifier.predict(X_test)
```

```
In [116... print(classification_report(y_test, y_pred))
```

	Top Used Banks			
	precision	recall	f1-score	support
-	0.00	0.00	0.00	1
Axis Bank	0.88	1.00	0.93	7
Bank Of Maharashtra	0.00	0.00	0.00	1
Canara Bank	1.00	1.00	1.00	1
DBS Bank	0.00	0.00	0.00	1
Federal Bank	0.00	0.00	0.00	1
HDFC Bank	0.84	1.00	0.91	26
ICICI Bank	1.00	1.00	1.00	19
IndusInd Bank	0.50	1.00	0.67	2
Kerala Bank	0.00	0.00	0.00	1
Kotak Mahindra Bank	0.96	1.00	0.98	25
Punjab National Bank	0.00	0.00	0.00	1
RBL Bank	0.00	0.00	0.00	1
State Bank Of India	1.00	0.67	0.80	3
Yes Bank	0.00	0.00	0.00	1
accuracy			0.90	91
macro avg	0.41	0.44	0.42	91
weighted avg	0.83	0.90	0.86	91

```
C:\Users\NITISH\anaconda3\Lib\site-packages\sklearn\metrics\_classification.py:146
9: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to
0.0 in labels with no predicted samples. Use `zero_division` parameter to control
this behavior.
```

```
_warn_prf(average, modifier, msg_start, len(result))
```

```
C:\Users\NITISH\anaconda3\Lib\site-packages\sklearn\metrics\_classification.py:146
9: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to
0.0 in labels with no predicted samples. Use `zero_division` parameter to control
this behavior.
```

```
_warn_prf(average, modifier, msg_start, len(result))
```

```
C:\Users\NITISH\anaconda3\Lib\site-packages\sklearn\metrics\_classification.py:146
9: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to
0.0 in labels with no predicted samples. Use `zero_division` parameter to control
this behavior.
```

```
_warn_prf(average, modifier, msg_start, len(result))
```

In [ ]:

```
In [117... subset = df[['Age group', 'Salary Acc', 'Primary Acc']]]
```

```
In [118... count_df = subset.groupby(['Age group', 'Salary Acc', 'Primary Acc']).size().reset_
```

```
In [119... import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

## MOST USED AS A SALARY ACC

```
In [120... import pandas as pd
import matplotlib.pyplot as plt

# Sample data
data = {'Salary Acc': ['Yes Bank', 'Bank Of Baroda', 'Axis Bank', 'None of the above']}

# Convert to DataFrame
df = pd.DataFrame(data)

# Filter out "None of the above" entries
```

```
df_filtered = df[df['Salary Acc'] != 'None of the above']

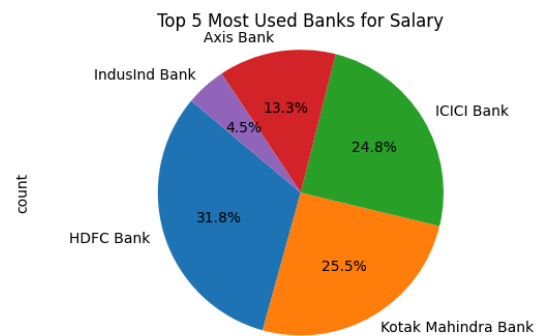
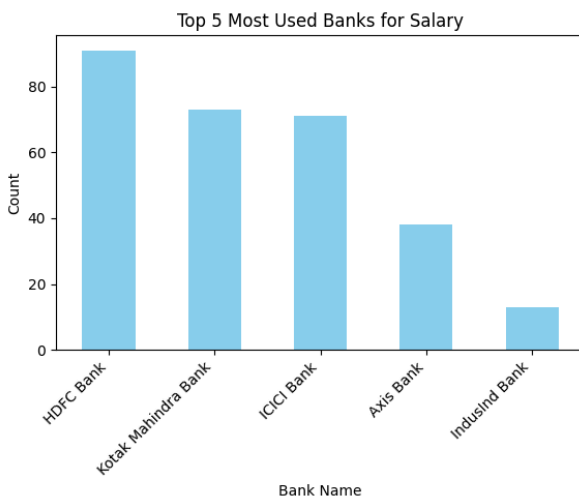
# Get top 5 banks
top_5_banks = df_filtered['Salary Acc'].value_counts().nlargest(5)

# Plot
plt.figure(figsize=(12, 5))

plt.subplot(1, 2, 1)
top_5_banks.plot(kind='bar', color='skyblue')
plt.title('Top 5 Most Used Banks for Salary')
plt.xlabel('Bank Name')
plt.ylabel('Count')
plt.xticks(rotation=45, ha='right')

plt.subplot(1, 2, 2)
top_5_banks.plot(kind='pie', autopct='%1.1f%', startangle=140)
plt.axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle.
plt.title('Top 5 Most Used Banks for Salary')

plt.tight_layout()
plt.show()
```



In [122...

```
import seaborn as sns
import matplotlib.pyplot as plt
from collections import Counter

# Remove 'None of the above' options and combine bank accounts
subset = subset[(subset[['Salary Acc', 'Primary Acc']] != 'None of the above').all()]
combined_accounts = subset['Salary Acc'].tolist() + subset['Primary Acc'].tolist()

# Count the frequency of each bank
bank_counts = Counter(combined_accounts)

# Sort banks and counts based on counts
banks_sorted, counts_sorted = zip(*sorted(bank_counts.items(), key=lambda x: x[1],

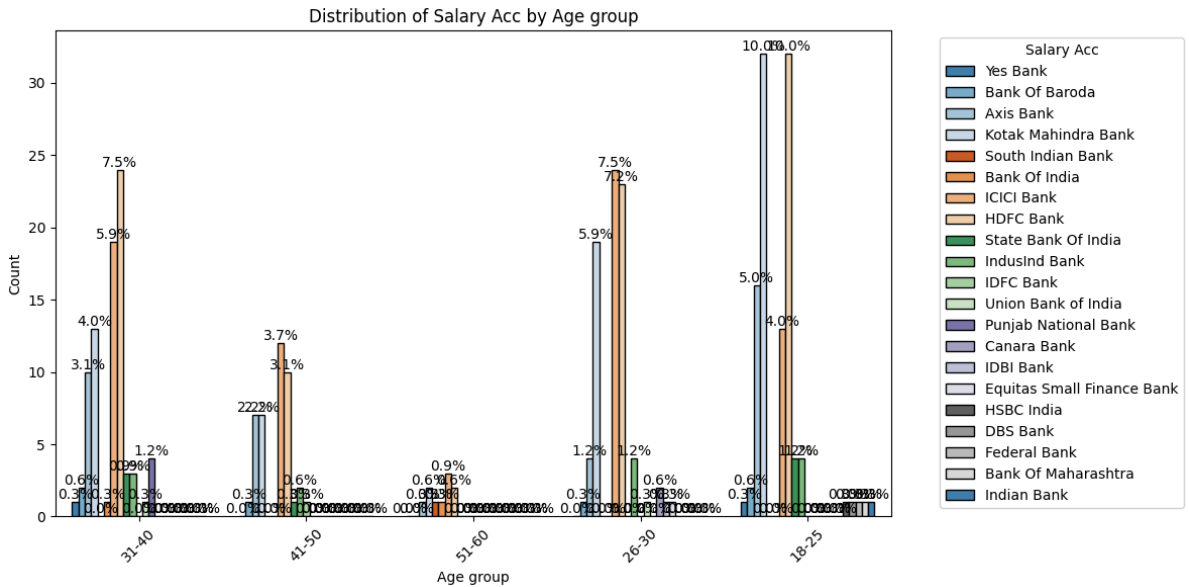
# Define custom colors for the bars
custom_colors = plt.cm.tab20c.colors[:len(banks_sorted)])

# Distribution of Salary Acc by Age group
plt.figure(figsize=(12, 6))
ax = sns.countplot(x='Age group', hue='Salary Acc', data=subset, palette=custom_col

# Add percentage labels above each bar
for p in ax.patches:
    percentage = '{:.1f}%'.format(100 * p.get_height() / len(subset))
```

```
ax.annotate(percentage, (p.get_x() + p.get_width() / 2, p.get_height()), ha='center', va='bottom', size=10, color='black')

plt.title('Distribution of Salary Acc by Age group')
plt.xlabel('Age group')
plt.ylabel('Count')
plt.legend(title='Salary Acc', bbox_to_anchor=(1.05, 1), loc='upper left')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```



In [123...

```
import pandas as pd

# Sample data
data = {'Salary Acc': ['Yes Bank', 'Bank Of Baroda', 'Axis Bank', 'None of the above']}

# Convert to DataFrame
df = pd.DataFrame(data)

print(len(df))
```

454

In [124...

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

# Sample data
data = {'Salary Acc': ['Yes Bank', 'Bank Of Baroda', 'Axis Bank', 'None of the above']}

# Convert to DataFrame
df = pd.DataFrame(data)

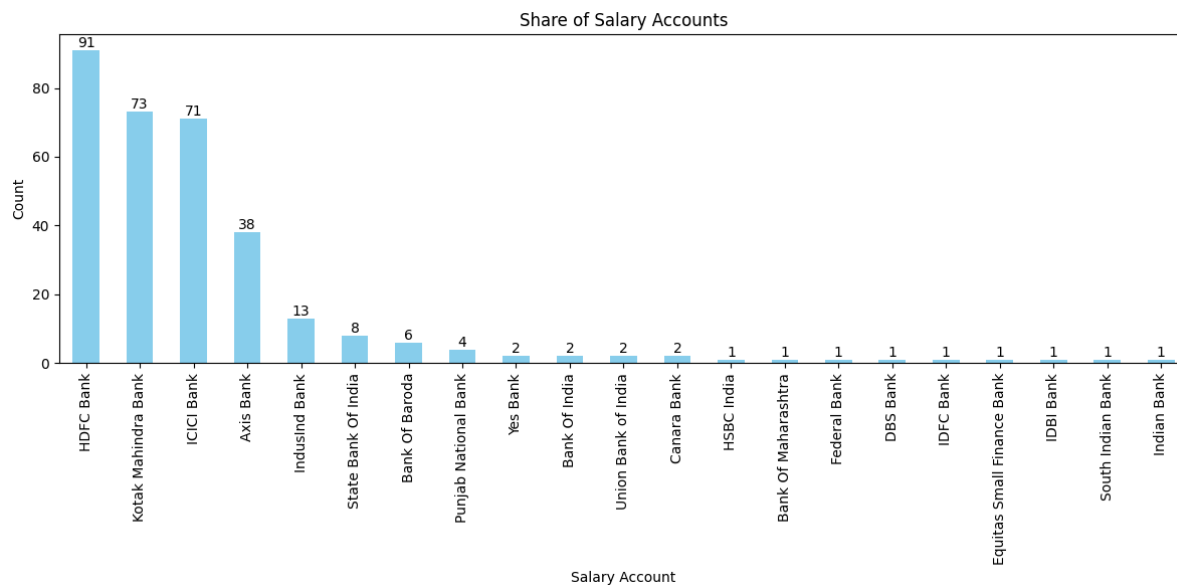
# Filter out "None of the above"
df_filtered = df[df['Salary Acc'] != 'None of the above']

# Plot
plt.figure(figsize=(12, 6))
df_filtered['Salary Acc'].value_counts().plot(kind='bar', color='skyblue')
plt.title('Share of Salary Accounts')
plt.xlabel('Salary Account')
plt.ylabel('Count')
plt.xticks(rotation=90)

# Display counts on each bar
counts = df_filtered['Salary Acc'].value_counts()
for i, count in enumerate(counts):
```

```
plt.text(i, count, str(count), ha='center', va='bottom')

plt.tight_layout()
plt.show()
```



## Primary Acc

In [125...

```
plt.figure(figsize=(8, 8))
df['Primary Acc'].value_counts().plot(kind='pie', autopct='%1.1f%%', startangle=140)
plt.title('Total Number of Primary Acc')
plt.ylabel('')
plt.show()
```

```
-----
KeyError                                Traceback (most recent call last)
File ~\anaconda3\Lib\site-packages\pandas\core\indexes\base.py:3805, in Index.get_loc(self, key)
    3804 try:
-> 3805     return self._engine.get_loc(casted_key)
    3806 except KeyError as err:

File index.pyx:167, in pandas._libs.index.IndexEngine.get_loc()

File index.pyx:196, in pandas._libs.index.IndexEngine.get_loc()

File pandas\_libs\hashtable_class_helper.pxi:7081, in pandas._libs.hashtable.PyObjectHashTable.get_item()

File pandas\_libs\hashtable_class_helper.pxi:7089, in pandas._libs.hashtable.PyObjectHashTable.get_item()

KeyError: 'Primary Acc'
```

The above exception was the direct cause of the following exception:

```
KeyError                                Traceback (most recent call last)
Cell In[125], line 2
      1 plt.figure(figsize=(8, 8))
----> 2 df['Primary Acc'].value_counts().plot(kind='pie', autopct='%1.1f%%', start
angle=140, colors=sns.color_palette('pastel'))
      3 plt.title('Total Number of Primary Acc')
      4 plt.ylabel('')

File ~\anaconda3\Lib\site-packages\pandas\core\frame.py:4090, in DataFrame.__getitem__(self, key)
    4088 if self.columns.nlevels > 1:
    4089     return self._getitem_multilevel(key)
-> 4090 indexer = self.columns.get_loc(key)
    4091 if is_integer(indexer):
    4092     indexer = [indexer]

File ~\anaconda3\Lib\site-packages\pandas\core\indexes\base.py:3812, in Index.get_loc(self, key)
    3807 if isinstance(casted_key, slice) or (
    3808     isinstance(casted_key, abc.Iterable)
    3809     and any(isinstance(x, slice) for x in casted_key)
    3810 ):
    3811     raise InvalidIndexError(key)
-> 3812     raise KeyError(key) from err
    3813 except TypeError:
    3814     # If we have a listlike key, _check_indexing_error will raise
    3815     # InvalidIndexError. Otherwise we fall through and re-raise
    3816     # the TypeError.
    3817     self._check_indexing_error(key)

KeyError: 'Primary Acc'
<Figure size 800x800 with 0 Axes>
```

In [126...

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

# Sample data
data = {'Primary Acc': ['Yes Bank', 'Bank Of Baroda', 'Axis Bank', 'None of the abc

# Convert to DataFrame
df = pd.DataFrame(data)
```

```

# Remove "None of the above" entries
df_filtered = df[df['Primary Acc'] != 'None of the above']

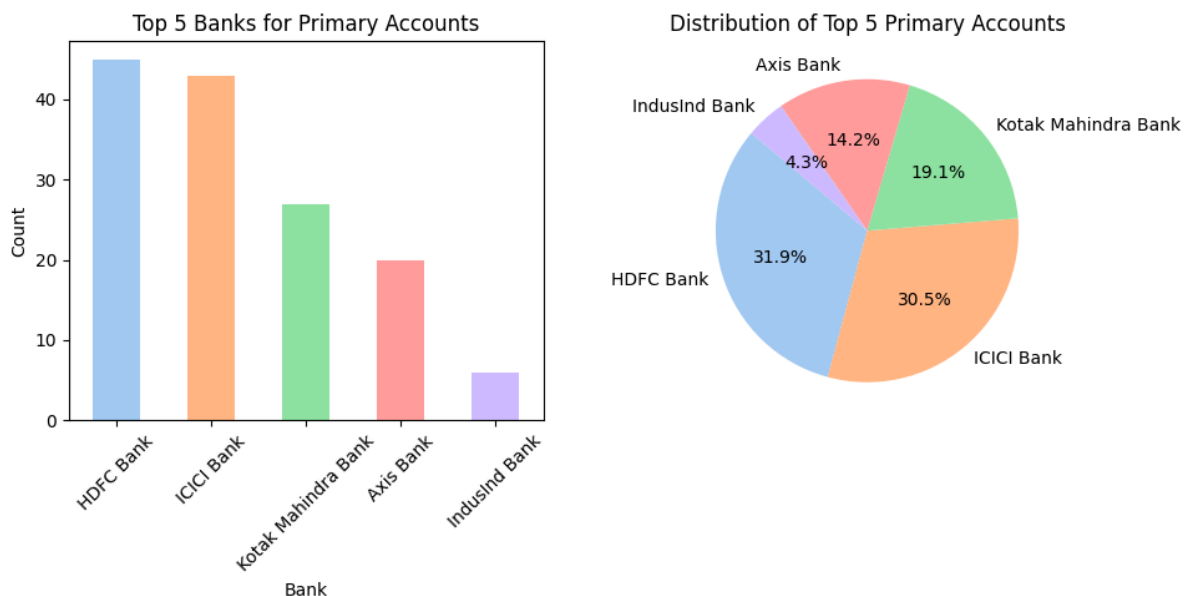
# Get top 5 banks
top_5_banks = df_filtered['Primary Acc'].value_counts().head(5)

# Plot bar chart for top 5 banks
plt.figure(figsize=(10, 5))
plt.subplot(1, 2, 1)
top_5_banks.plot(kind='bar', color=sns.color_palette('pastel'))
plt.title('Top 5 Banks for Primary Accounts')
plt.xlabel('Bank')
plt.ylabel('Count')
plt.xticks(rotation=45)

# Plot pie chart for top 5 banks
plt.subplot(1, 2, 2)
top_5_banks.plot(kind='pie', autopct='%1.1f%%', startangle=140, colors=sns.color_palette('pastel'))
plt.title('Distribution of Top 5 Primary Accounts')
plt.ylabel('')

plt.tight_layout()
plt.show()

```



In [127...

```

import pandas as pd
import matplotlib.pyplot as plt

# Sample data
data = {'Primary Acc': ['Yes Bank', 'HDFC Bank', 'Kotak Mahindra Bank', 'ICICI Bank', 'IndusInd Bank']}

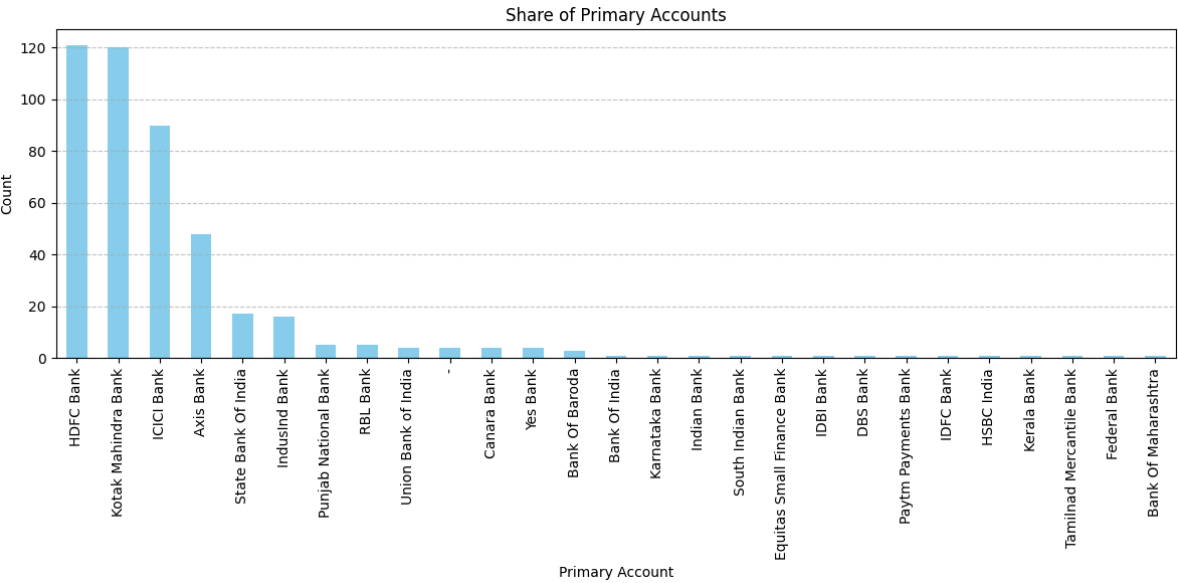
# Convert to DataFrame
df = pd.DataFrame(data)

# Count occurrences of each primary account
primary_acc_counts = df['Primary Acc'].value_counts()

# Plot
plt.figure(figsize=(12, 6))
primary_acc_counts.plot(kind='bar', color='skyblue')
plt.title('Share of Primary Accounts')
plt.xlabel('Primary Account')
plt.ylabel('Count')
plt.xticks(rotation=90)

```

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



```
In [128... df=pd.read_csv("C:/Users/NITISH/OneDrive/Documents/Desktop/Bank Data/TOP BANKS.csv")
```

```
In [129... print(df)
```



	Gender	City	Age group	Occupation	BANK1 \
0	Male	Mumbai	31-40	Own Business	HDFC Bank
1	Female	Bangalore	41-50	Salaried ☑ Private	ICICI Bank
2	Male	Ahmedabad	41-50	Own Business	Kotak Mahindra Bank
3	Male	Bangalore	26-30	Salaried ☑ Private	HDFC Bank
4	Male	Ahmedabad	51-60	Own Business	Kotak Mahindra Bank
..	...	...	...	...	...
450	Male	Bangalore	18-25	Salaried ☑ Private	Kotak Mahindra Bank
451	Male	Bangalore	18-25	Salaried ☑ Private	IndusInd Bank
452	Male	Delhi	18-25	Salaried ☑ Private	HDFC Bank
453	Male	Ahmedabad	18-25	Salaried ☑ Private	Kotak Mahindra Bank
454	Male	Bangalore	18-25	Salaried ☑ Private	Kotak Mahindra Bank

	BANK2	BANK3	Kotak Mahindra Bank \
0	ICICI Bank	Yes Bank	Kotak Mahindra Bank
1	HDFC Bank	Kotak Mahindra Bank	Kotak Mahindra Bank
2	HDFC Bank	Axis Bank	-
3	ICICI Bank	Kotak Mahindra Bank	-
4	HDFC Bank	State Bank Of India	Kotak Mahindra Bank
..	...	...	...
450	IndusInd Bank	HDFC Bank	Kotak Mahindra Bank
451	Indian Bank	Federal Bank	-
452	Kotak Mahindra Bank	ICICI Bank	-
453	ICICI Bank	Central Bank Of India	Kotak Mahindra Bank
454	State Bank Of India	NaN	Kotak Mahindra Bank

	HDFC Bank	ICICI Bank	Axis Bank	IndusInd Bank	RBL Bank \
0	-	-	-	-	-
1	-	-	-	-	-
2	-	-	-	IndusInd Bank	-
3	HDFC Bank	-	-	-	-
4	-	-	-	-	-
..	...	...	...	...	...
450	-	-	-	-	-
451	-	-	-	IndusInd Bank	-
452	HDFC Bank	-	-	-	-
453	-	-	-	-	-
454	-	-	-	-	-

	Other 1	Other 2	Salary Acc	Primary Acc
0	-	-	Yes Bank	Yes Bank
1	-	-	Bank Of Baroda	HDFC Bank
2	South Indian Bank	-	Axis Bank	Kotak Mahindra Bank
3	Bank Of India	-	None of the above	ICICI Bank
4	-	-	Kotak Mahindra Bank	Kotak Mahindra Bank
..	...	...	...	...
450	-	-	Kotak Mahindra Bank	Kotak Mahindra Bank
451	Indian Bank	-	Indian Bank	IndusInd Bank
452	-	-	HDFC Bank	HDFC Bank
453	-	-	ICICI Bank	Kotak Mahindra Bank
454	State Bank Of India	-	None of the above	Kotak Mahindra Bank

[455 rows x 17 columns]

## Total Number of (Primary+Salary) Bank Users

In [130...

```
import matplotlib.pyplot as plt
from collections import Counter

# Extract salary accounts and primary accounts
```

```

salary_accounts = df.loc[df['Salary Acc'] != 'None of the above', 'Salary Acc'].tol
primary_accounts = df.loc[df['Primary Acc'] != 'None of the above', 'Primary Acc'].

# Combine the Lists
combined_accounts = salary_accounts + primary_accounts

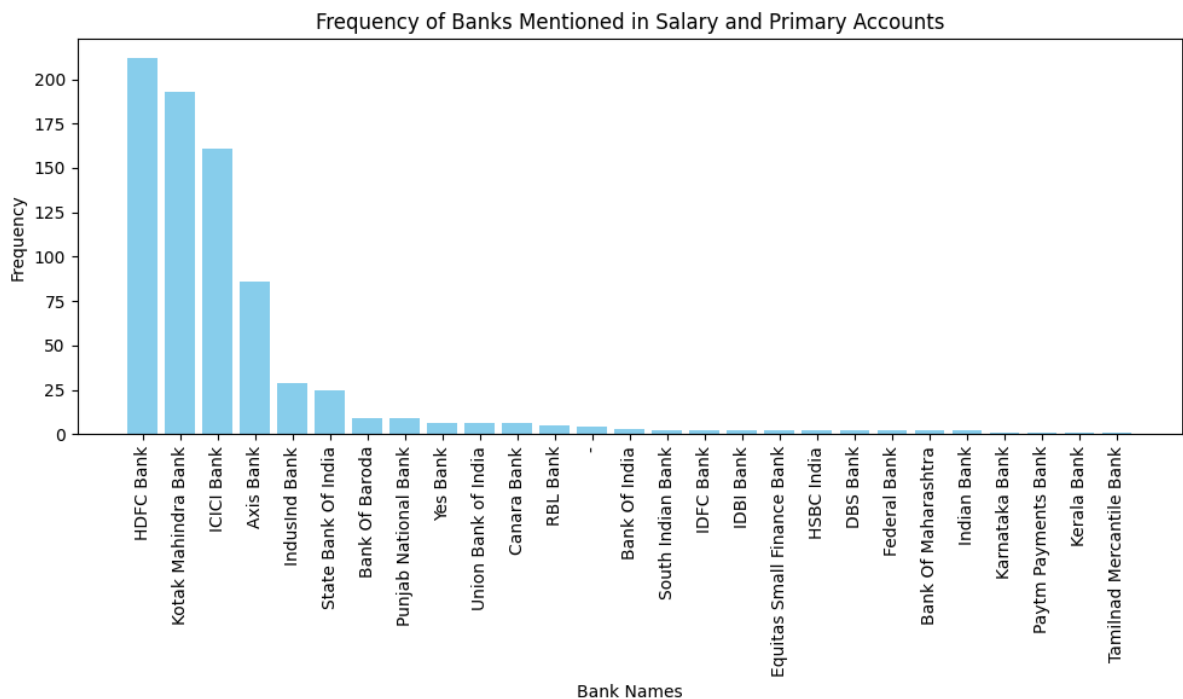
# Count the frequency of each bank
bank_counts = Counter(combined_accounts)

# Extract bank names and corresponding counts
banks = list(bank_counts.keys())
counts = list(bank_counts.values())

# Sort banks and counts based on counts
sorted_indices = sorted(range(len(counts)), key=lambda k: counts[k], reverse=True)
banks_sorted = [banks[i] for i in sorted_indices]
counts_sorted = [counts[i] for i in sorted_indices]

# Plotting
plt.figure(figsize=(10, 6))
plt.bar(banks_sorted, counts_sorted, color='skyblue')
plt.xlabel('Bank Names')
plt.ylabel('Frequency')
plt.title('Frequency of Banks Mentioned in Salary and Primary Accounts')
plt.xticks(rotation=90)
plt.tight_layout()
plt.show()

```



In [ ]: