▼ Personal Transaction Analysis

▼ HDFC Bank Account

```
# load the required Libraries
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
# Initilise the drive for dataset
from google.colab import drive
# initilise tyhe path
drive.mount('/content/drive')
# Setting Address
import os
os.chdir('/content/drive/My Drive/dummy folder')
Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).
# Load the data (HDFC bank Statement)
# This data is from 1st April 2022 to 31st March 2023
# Load the data
df_HDFC = pd.read_excel("my bank statement.xlsx")
df_HDFC.head()
```

| | Date | Narration | Withdrawal Amt. | Deposit Amt. | Closing Balance | Month | Year |
|---|------------|--|-----------------|--------------|-----------------|-------|------|
| 0 | 2022-04-04 | UPI-VISHAL VINAYAK BHOYA-VISHALBHOYAR313 | NaN | 200.0 | 442.19 | 4 | 2022 |
| 1 | 2022-04-07 | UPI-ANIL NARBAT BISEN-ANILBISEN229@OKSBI | NaN | 50.0 | 492.19 | 4 | 2022 |
| 2 | 2022-04-07 | UPI-SHRIRAM GENERAL STOR-GPAY-1119808673 | 50.0 | NaN | 442.19 | 4 | 2022 |
| 3 | 2022-04-12 | UPI-IDRISH SHHA-BHARATPE.9051322211@FBP | 30.0 | NaN | 412.19 | 4 | 2022 |
| 4 | 2022-04-13 | UPI-RAMESHWAR VASANTRAO | NaN | 100.0 | 512.19 | 4 | 2022 |

```
# Convert 'Date' column to datetime data type
df_HDFC['Date'] = pd.to_datetime(df_HDFC['Date'])

# Handling missing values (if any)
df_HDFC.fillna(0, inplace=True) # Replace NaN with 0 for Withdrawal and Deposit Amounts

# Drop unnecessary columns (if needed)
df_HDFC.drop(['Month', 'Year'], axis=1, inplace=True)
```

 $df_HDFC.head()$

| | Date | Narration | Withdrawal Amt. | Deposit Amt. | Closing Balance | 1 | ılı |
|---|------------|--|-----------------|--------------|-----------------|---|-----|
| 0 | 2022-04-04 | UPI-VISHAL VINAYAK BHOYA-VISHALBHOYAR313 | 0.0 | 200.0 | 442.19 | | |
| 1 | 2022-04-07 | UPI-ANIL NARBAT BISEN-ANILBISEN229@OKSBI | 0.0 | 50.0 | 492.19 | | |
| 2 | 2022-04-07 | UPI-SHRIRAM GENERAL STOR-GPAY-1119808673 | 50.0 | 0.0 | 442.19 | | |
| 3 | 2022-04-12 | UPI-IDRISH SHHA-BHARATPE.9051322211@FBP | 30.0 | 0.0 | 412.19 | | |
| 4 | 2022-04-13 | UPI-RAMESHWAR VASANTRAO | 0.0 | 100.0 | 512.19 | | |
| | | | | | | | |

```
# Total income and total expenses
total_income = df_HDFC['Deposit Amt.'].sum()
total_expenses = df_HDFC['Withdrawal Amt.'].sum()
# Print those
print("total_income: ",total_income)
```

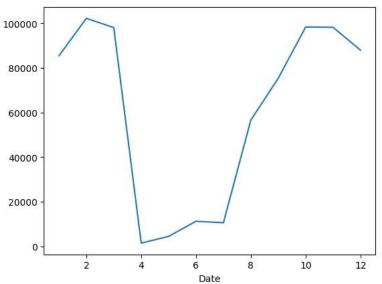
```
print("total_expenses: ",total_expenses )
```

total_income: 729578.0
total_expenses: 729503.33

Monthly spending patterns

monthly_expenses = df_HDFC.groupby(df_HDFC['Date'].dt.month)['Withdrawal Amt.'].sum()
monthly_expenses.plot()

<Axes: xlabel='Date'>



Major expenses

top_expenses = df_HDFC.nlargest(5, 'Withdrawal Amt.')

top_expenses

| | Date | Narration | Withdrawal Amt. | Deposit Amt. | Closing Balance | 1 | ılı |
|----|----------------------|--|-----------------|--------------|-----------------|---|-----|
| 50 | 63 2022-11-11 | UPI-VISHAL VINAYAK BHOYA-VISHALBHOYAR313 | 50000.0 | 0.0 | 18543.57 | | |
| 6 | 56 2022-12-02 | UPI-VISHAL VINAYAK BHOYA-VISHALBHOYAR313 | 50000.0 | 0.0 | 2765.81 | | |
| 79 | 90 2023-01-04 | UPI-VISHAL VINAYAK BHOYA-VISHALBHOYAR313 | 50000.0 | 0.0 | 4087.80 | | |
| 94 | 45 2023-02-13 | UPI-VISHAL VINAYAK BHOYA-VISHALBHOYAR313 | 50000.0 | 0.0 | 12579.29 | | |
| 10 | 28 2023-03-02 | UPI-VISHAL VINAYAK BHOYA-VISHALBHOYAR313 | 50000.0 | 0.0 | 3888.27 | | |

Average Monthly Balance:

Calculate the average monthly balance based on the closing balance for each month.

df_HDFC['Month'] = df_HDFC['Date'].dt.month

average_monthly_balance = df_HDFC.groupby('Month')['Closing Balance'].mean()

average_monthly_balance.plot(kind = "bar")

plt.show()

```
# Income vs. Expenses Over Time:

# Visualize the trend of your income and expenses over the months.

monthly_income = df_HDFC.groupby('Month')['Deposit Amt.'].sum()

monthly_expenses = df_HDFC.groupby('Month')['Withdrawal Amt.'].sum()

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

plt.plot(monthly_income.index, monthly_income, label='Income', marker='o')

plt.plot(monthly_expenses.index, monthly_expenses, label='Expenses', marker='o')

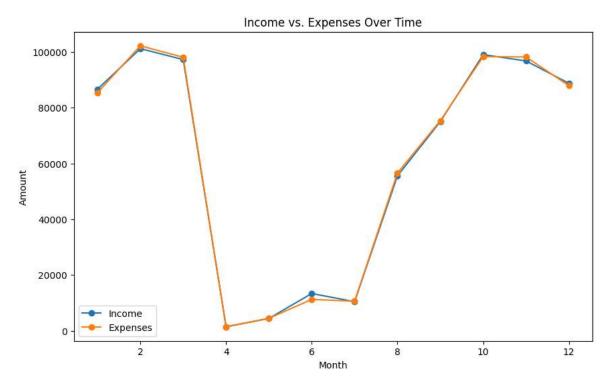
plt.xlabel('Month')

plt.ylabel('Amount')

plt.title('Income vs. Expenses Over Time')

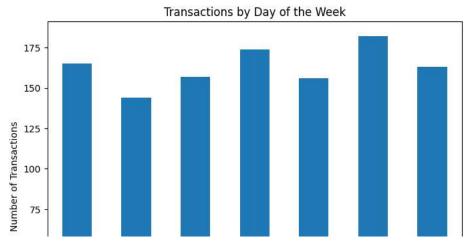
plt.legend()

plt.show()
```



```
# Day of the Week Analysis:
# Check if there's any pattern in your transactions based on the day of the week.
df_HDFC['Day_of_Week'] = df_HDFC['Date'].dt.day_name()
transactions_by_day = df_HDFC.groupby('Day_of_Week').size()

plt.figure(figsize=(8, 6))
transactions_by_day.plot(kind='bar')
plt.xlabel('Day of the Week')
plt.ylabel('Number of Transactions')
plt.title('Transactions by Day of the Week')
plt.show()
```



Top Income Sources:

If the 'Narration' column contains details about income sources, you can identify the top sources.

top_income_sources = df_HDFC.nlargest(5, 'Deposit Amt.')

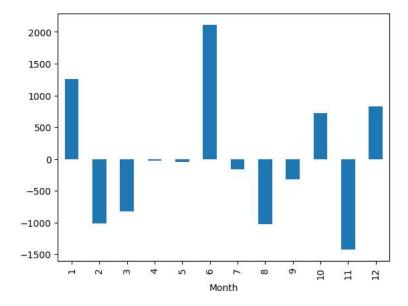
top_income_sources

| | Date | Narration | Withdrawal Amt. | Deposit Amt. | Closing Balance | Month | Day_of_Week | 1 |
|------|------------|--|-----------------|--------------|-----------------|-------|-------------|---|
| 1027 | 2023-03-02 | UPI-AMAN JANGID-PAYTQR6377020338@PAYTM-B | 0.0 | 53527.0 | 53888.27 | 3 | Thursday | |
| 789 | 2023-01-04 | UPI-AMAN JANGID-PAYTQR6377020338@PAYTM-B | 0.0 | 53500.0 | 54087.80 | 1 | Wednesday | |
| 898 | 2023-02-02 | UPI-AMAN JANGID-PAYTQR6377020338@PAYTM-B | 0.0 | 52858.0 | 54798.80 | 2 | Thursday | |
| 507 | 2022-11-01 | UPI-AMAN JANGID-PAYTQR6377020338@PAYTM-B | 0.0 | 52650.0 | 54195.57 | 11 | Tuesday | |
| 653 | 2022-12-02 | IMPS-233611598080-AMAN JANGID-PYTM-XXXXX | 0.0 | 52195.0 | 52815.81 | 12 | Fridav | |



Calculate the net cash flow for each month (income minus expenses). df_HDFC['Net_Cash_Flow'] = df_HDFC['Deposit Amt.'] - df_HDFC['Withdrawal Amt.'] monthly_cash_flow = df_HDFC.groupby('Month')['Net_Cash_Flow'].sum() monthly_cash_flow.plot(kind = "bar")

plt.show()



```
# Transaction Frequency Analysis:
```

Analyze the frequency of transactions during the year.

transaction_frequency = df_HDFC.groupby('Month').size()

plt.figure(figsize=(8, 6))

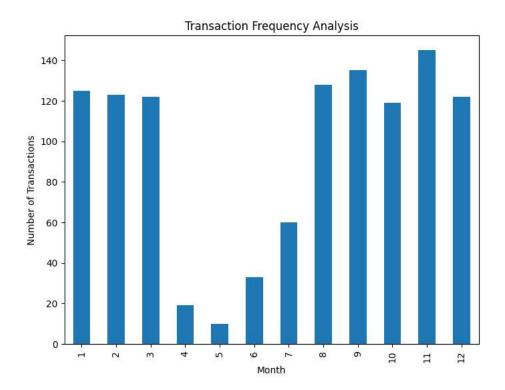
transaction_frequency.plot(kind='bar')

plt.xlabel('Month')

plt.ylabel('Number of Transactions')

plt.title('Transaction Frequency Analysis')

plt.show()

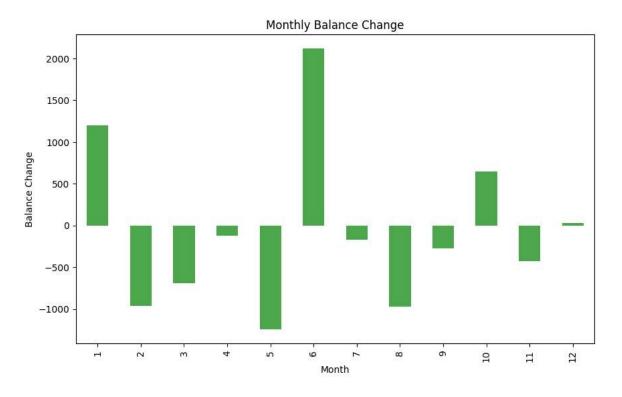


```
# Expenses by Day of the Week:
# Analyze your expenses on each day of the week to identify spending patterns.
expenses_by_day_of_week = df_HDFC[df_HDFC['Withdrawal Amt.'] > 0].groupby('Day_of_Week')['Withdrawal Amt.'].sum()

plt.figure(figsize=(8, 6))
expenses_by_day_of_week.plot(kind='bar')
plt.xlabel('Day of the Week')
plt.ylabel('Total Expenses')
plt.title('Expenses by Day of the Week')
plt.show()
```

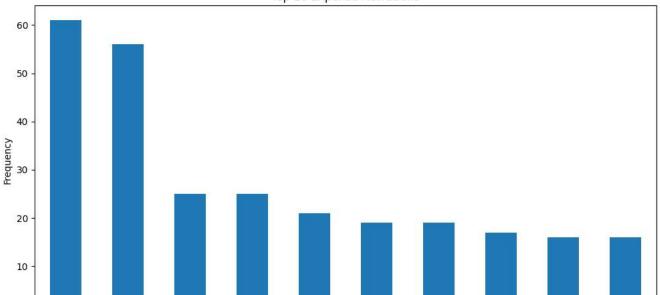
```
# Monthly Balance Change:
# Plot the change in your account balance over the months to see how it fluctuates.
monthly_balance_change = df_HDFC.groupby('Month')['Closing Balance'].last() - df_HDFC.groupby('Month')['Closing Balance'].first()

plt.figure(figsize=(10, 6))
monthly_balance_change.plot(kind='bar', color='g', alpha=0.7)
plt.xlabel('Month')
plt.ylabel('Balance Change')
plt.title('Monthly Balance Change')
plt.show()
```

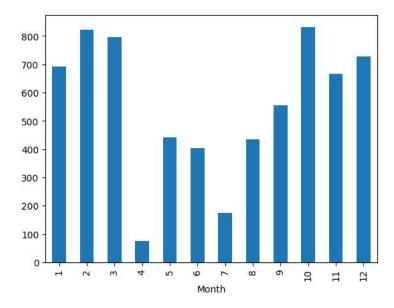


```
# Expense Distribution by Narration:
# Analyze the distribution of expenses based on the descriptions in the 'Narration' column.
plt.figure(figsize=(12, 6))
top_narrations = df_HDFC['Narration'].value_counts().nlargest(10)
top_narrations.plot(kind='bar')
plt.xlabel('Narration')
plt.ylabel('Narration')
plt.ylabel('Frequency')
plt.title('Top 10 Expense Narrations')
plt.show()
```

Top 10 Expense Narrations



Average Withdrawal and Deposit Amounts:
Calculate the average withdrawal and deposit amounts for each month.
average_withdrawal_per_month = df_HDFC.groupby('Month')['Withdrawal Amt.'].mean()
average_deposit_per_month = df_HDFC.groupby('Month')['Deposit Amt.'].mean()
average_deposit_per_month.plot(kind = "bar")
plt.show()



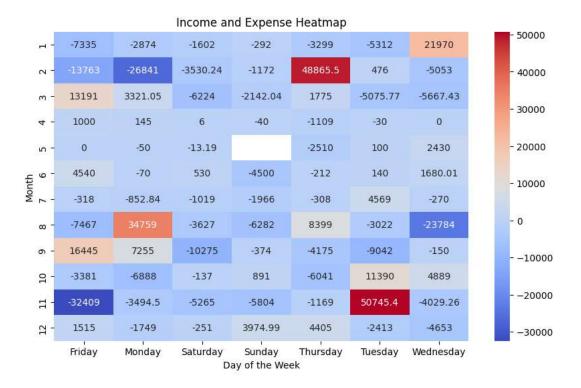
average_withdrawal_per_month.plot(kind = "line")
plt.show()

```
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```

Income and Expense Heatmap:

Create a heatmap to visualize the distribution of income and expenses across different months. income_expense_heatmap = df_HDFC.pivot_table(index='Month', columns='Day_of_Week', values='Net_Cash_Flow', aggfunc='sum')

```
plt.figure(figsize=(10, 6))
sns.heatmap(income_expense_heatmap, cmap='coolwarm', annot=True, fmt='g')
plt.xlabel('Day of the Week')
plt.ylabel('Month')
plt.title('Income and Expense Heatmap')
plt.show()
```

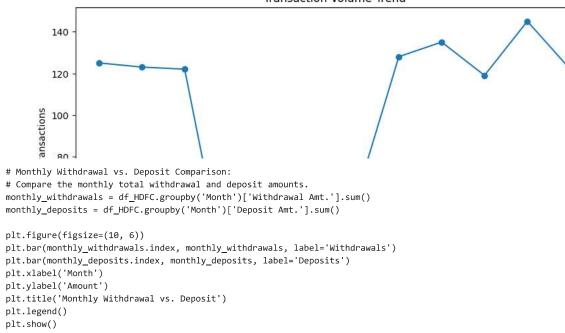


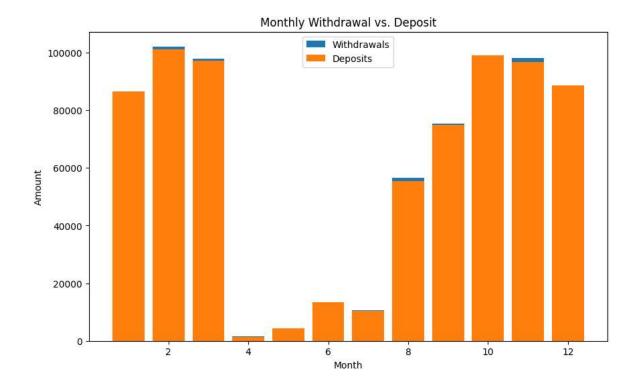
```
# Transaction Volume Trend:
```

Visualize the trend of the number of transactions over the months.
monthly_transaction_count = df_HDFC.groupby('Month').size()

```
plt.figure(figsize=(10, 6))
monthly_transaction_count.plot(kind='line', marker='o')
plt.xlabel('Month')
plt.ylabel('Number of Transactions')
plt.title('Transaction Volume Trend')
plt.show()
```

Transaction Volume Trend





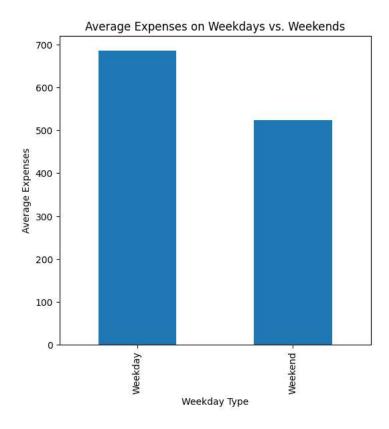
bank_statement_df = df_HDFC

```
# Average Daily Expenses:
# Calculate the average daily expenses for each month.
bank_statement_df['Day'] = bank_statement_df['Date'].dt.day
daily_expenses = bank_statement_df.groupby(['Month', 'Day'])['Withdrawal Amt.'].sum().reset_index()
average_daily_expenses = daily_expenses.groupby('Month')['Withdrawal Amt.'].mean()

# Weekday vs. Weekend Expenses:
# Compare your average expenses on weekdays and weekends.
def categorize_weekday(weekday):
    if weekday in ['Saturday', 'Sunday']:
        return 'Weekend'
    return 'Weekday'
```

```
bank_statement_df['Weekday_Type'] = bank_statement_df['Day_of_Week'].apply(categorize_weekday)
expenses_by_weekday_type = bank_statement_df.groupby('Weekday_Type')['Withdrawal Amt.'].mean()

plt.figure(figsize=(6, 6))
expenses_by_weekday_type.plot(kind='bar')
plt.xlabel('Weekday Type')
plt.ylabel('Average Expenses')
plt.title('Average Expenses on Weekdays vs. Weekends')
plt.show()
```



```
# Transaction Descriptions Word Cloud:
# Create a word cloud to visualize the most frequent words in transaction descriptions.
from wordcloud import WordCloud

narration_words = ' '.join(bank_statement_df['Narration'].astype(str))
wordcloud = WordCloud(width=800, height=400, background_color='white').generate(narration_words)

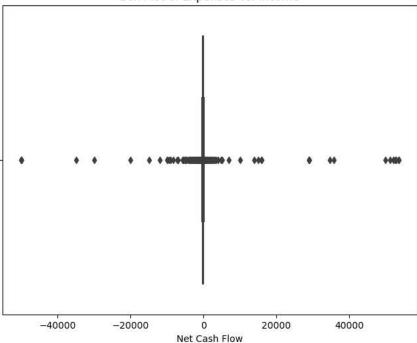
plt.figure(figsize=(10, 6))
plt.imshow(wordcloud, interpolation='bilinear')
plt.axis('off')
plt.title('Transaction Descriptions Word Cloud')
plt.show()
```

Transaction Descriptions Word Cloud



```
# Expenses vs. Income Box Plot:
# Compare the distribution of expenses and income using a box plot.
plt.figure(figsize=(8, 6))
sns.boxplot(x='Net_Cash_Flow', data=bank_statement_df)
plt.xlabel('Net Cash Flow')
plt.title('Box Plot of Expenses vs. Income')
plt.show()
```

Box Plot of Expenses vs. Income



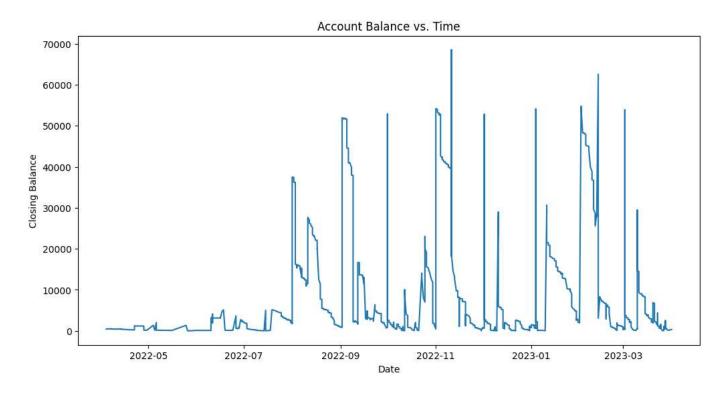
```
# Expenses vs. Closing Balance Scatter Plot:
# Visualize the relationship between expenses and the closing balance for each transaction.
plt.figure(figsize=(10, 6))
plt.scatter(bank_statement_df['Withdrawal Amt.'], bank_statement_df['Closing Balance'], alpha=0.7)
plt.xlabel('Expenses')
plt.ylabel('Closing Balance')
plt.title('Expenses vs. Closing Balance')
plt.show()
```

Expenses vs. Closing Balance

```
70000
# Monthly Savings:
# Calculate your monthly savings by subtracting total expenses from total income.
monthly_savings = monthly_income - monthly_expenses
# Biggest Income and Expense Days:
# Identify the days with the highest income and expenses.
biggest_income_day = bank_statement_df.nlargest(1, 'Deposit Amt.')
biggest_expense_day = bank_statement_df.nlargest(1, 'Withdrawal Amt.')
# Withdrawal and Deposit Frequency:
# Analyze the frequency of withdrawals and deposits over the months.
with drawal\_frequency = bank\_statement\_df[bank\_statement\_df['Withdrawal Amt.'] > 0].groupby('Month').size() \\
deposit_frequency = bank_statement_df[bank_statement_df['Deposit Amt.'] > 0].groupby('Month').size()
plt.figure(figsize=(10, 6))
withdrawal_frequency.plot(kind='line', marker='o', label='Withdrawals')
deposit_frequency.plot(kind='line', marker='o', label='Deposits')
plt.xlabel('Month')
plt.ylabel('Number of Transactions')
plt.title('Withdrawal and Deposit Frequency')
plt.legend()
plt.show()
```

Withdrawal and Deposit Frequency Withdrawals Deposits 100 40 20 40 Month

```
# Balance vs. Time Plot:
# Visualize the change in your account balance over time.
plt.figure(figsize=(12, 6))
plt.plot(bank_statement_df['Date'], bank_statement_df['Closing Balance'])
plt.xlabel('Date')
plt.ylabel('Closing Balance')
plt.title('Account Balance vs. Time')
plt.show()
```



```
# Expense vs. Income Ratio:
# Calculate the ratio of total expenses to total income for each month.
expense_income_ratio = total_expenses / total_income

# Expenses by Day of the Month:
# Analyze your expenses on different days of the month.
expenses_by_day_of_month = bank_statement_df.groupby(bank_statement_df['Date'].dt.day)['Withdrawal Amt.'].sum()

plt.figure(figsize=(10, 6))
plt.plot(expenses_by_day_of_month.index, expenses_by_day_of_month, marker='o')
plt.xlabel('Day of the Month')
plt.ylabel('Expenses')
plt.title('Expenses by Day of the Month')
plt.show()
```

Expenses by Day of the Month

```
# Deposit and Withdrawal Trends Over Time:

# Visualize the trends of deposits and withdrawals over time.

plt.figure(figsize=(12, 6))

plt.plot(bank_statement_df['Date'], bank_statement_df['Deposit Amt.'], label='Deposits')

plt.plot(bank_statement_df['Date'], bank_statement_df['Withdrawal Amt.'], label='Withdrawals')

plt.xlabel('Date')

plt.ylabel('Amount')

plt.title('Deposit and Withdrawal Trends Over Time')

plt.legend()

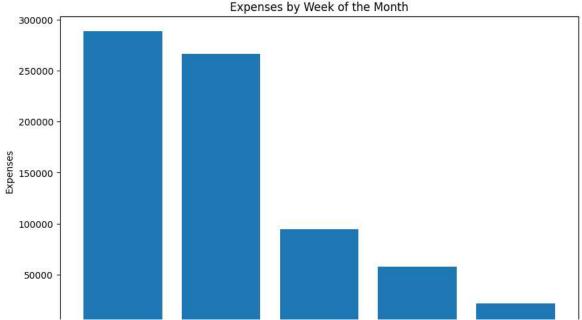
plt.show()
```

Deposit and Withdrawal Trends Over Time Deposits Withdrawals 50000 40000 30000 Amount 20000 10000 0 2022-05 2022-07 2022-09 2022-11 2023-01 2023-03 Date

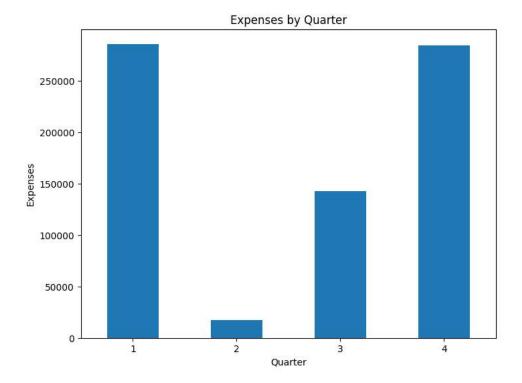
```
# Average Expense per Transaction:
# Calculate the average expense per transaction for each month.
bank_statement_df['Transaction_Amt'] = bank_statement_df['Withdrawal Amt.'] * -1
average_expense_per_transaction = bank_statement_df.groupby('Month')['Transaction_Amt'].mean()

# Expenses by Week of the Month:
# # Analyze your expenses based on the week of the month (e.g., 1st week, 2nd week, etc.).
week_of_month = bank_statement_df['Date'].apply(lambda x: (x.day - 1) // 7 + 1)
expenses_by_week_of_month = bank_statement_df.groupby(week_of_month)['Withdrawal Amt.'].sum()

plt.figure(figsize=(10, 6))
plt.bar(expenses_by_week_of_month.index, expenses_by_week_of_month)
plt.xlabel('Week of the Month')
plt.ylabel('Expenses')
plt.title('Expenses by Week of the Month')
plt.show()
```



```
# Expenses by Quarter:
# Analyze your expenses on a quarterly basis.
quarterly_expenses = bank_statement_df.groupby(bank_statement_df['Date'].dt.quarter)['Withdrawal Amt.'].sum()
plt.figure(figsize=(8, 6))
quarterly_expenses.plot(kind='bar', rot=0)
plt.xlabel('Quarter')
plt.ylabel('Expenses')
plt.title('Expenses by Quarter')
plt.show()
```



```
# Financial Ratios:
# Calculate financial ratios to assess your financial health, such as the Debt-to-Income Ratio, Savings Ratio, or Expense-to-Income Ratio.
# Debt-to-Income Ratio: Total Monthly Debt Payments / Monthly Gross Income
debt_to_income_ratio = total_expenses / monthly_income
# Savings Ratio: Monthly Savings / Monthly Income
```

savings_ratio = monthly_savings / monthly_income

```
# Expense-to-Income Ratio: Total Monthly Expenses / Monthly Gross Income
expense_to_income_ratio = total_expenses / monthly_income
```

from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler

Select relevant features for clustering (e.g., 'Withdrawal Amt.' and 'Deposit Amt.')
features_for_clustering = bank_statement_df[['Withdrawal Amt.', 'Deposit Amt.']]

Standardize the features
scaler = StandardScaler()

scaled_features = scaler.fit_transform(features_for_clustering)

Determine the optimal number of clusters using the Elbow method or other techniques
kmeans = KMeans(n_clusters=3) # You can choose the appropriate number of clusters
bank_statement_df['Cluster'] = kmeans.fit_predict(scaled_features)

/usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from 10 warnings.warn(



bank_statement_df.head()

| | Date | Narration | Withdrawal Amt. | Deposit Amt. | Closing Balance | Month | Day_of_Week | Net_Cash_Flow | Day | Weekday_Type | Transaction_/ |
|---|-------------------------|--|--------------------|-----------------|--------------------|-------|-------------|---------------|-----|--------------|---------------|
| 0 | 2022- 04-04 | UPI-VISHAL VINAYAK BHOYA- VISHALBHOYAR313 | 0.0 | 200.0 | 442.19 | 4 | Monday | 200.0 | 4 | Weekday | - |
| 1 | 2022- 04-07 | UPI-ANIL NARBAT BISEN- ANILBISEN229@OKSBI | 0.0 | 50.0 | 492.19 | 4 | Thursday | 50.0 | 7 | Weekday | - |
| 2 | 2022- 04 - 07 | UPI-SHRIRAM GENERAL STOR-GPAY-1119808673 | 50.0 | 0.0 | 442.19 | 4 | Thursday | -50.0 | 7 | Weekday | -5 |
| 3 | 2022- 04-12 | UPI-IDRISH SHHA- BHARATPE.9051322211@FBP | 30.0 | 0.0 | 412.19 | 4 | Tuesday | -30.0 | 12 | Weekday | -3 |
| 4 | 2022 - 04-13 | UPI-RAMESHWAR VASANTRAO | 0.0 | 100.0 | 512.19 | 4 | Wednesday | 100.0 | 13 | Weekday | - |





