## Author: Part 1: Mohit Bansal (Manager, Advisory, PwC India)



### Section 1: Import Data and read the file

```
#**Connect Google Colab to My google drive**
from google.colab import drive
drive.mount('/content/drive')
    Mounted at /content/drive
```

```
%matplotlib inline
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
bank_1 = pd.read_csv('/content/drive/MyDrive/Bank_Data/Bank_Data_v1.csv')
bank_1['month_year'] = pd.to_datetime(bank_1['post_date']).dt.to_period('M')
print(bank_1)
```

|       | Lead ID | bankid | <br>Primary_key               | month_year |
|-------|---------|--------|-------------------------------|------------|
| 0     | 308148  | 8535   | <br>308148_8535_12460_42439_1 | 2016-03    |
| 1     | 308148  | 8535   | <br>308148_8535_12460_42450_1 | 2016-03    |
| 2     | 308148  | 8535   | <br>308148_8535_12460_42450_3 | 2016-03    |
| 3     | 308148  | 8535   | <br>308148_8535_12460_42450_2 | 2016-03    |
| 4     | 308148  | 8535   | <br>308148_8535_12460_42451_1 | 2016-03    |
|       |         |        | <br>•••                       |            |
| 29024 | 330698  | 8545   | <br>330698_8545_14374_42772_7 | 2017-02    |
| 29025 | 330698  | 8545   | <br>330698_8545_14374_42773_2 | 2017-02    |
| 29026 | 330698  | 8545   | <br>330698_8545_14374_42773_4 | 2017-02    |
| 29027 | 330698  | 8545   | <br>330698_8545_14374_42773_3 | 2017-02    |
| 29028 | 330698  | 8545   | <br>330698_8545_14374_42773_1 | 2017-02    |

bank\_1.head(5)

|   | Lead<br>ID | bankid | bank_account_id | account_number | Industry                              | post_date | descripti  |
|---|------------|--------|-----------------|----------------|---------------------------------------|-----------|--|
| 0 | 308148     | 8535   | 12460           | xxxx9928       | Accommodation and Food Services       | 10-Mar-16 | DEPOSIT<br>NUMB<br>xx68                            |
| 1 | 308148     | 8535   | 12460           | xxxx9928       | Accommodation<br>and Food<br>Services | 21-Mar-16 | ATM CA<br>DEPOS<br>03/21 2904<br>BELT LI<br>RD IR' |
| 2 | 308148     | 8535   | 12460           | xxxx9928       | Accommodation and Food Services       | 21-Mar-16 | ATM CA<br>DEPOS<br>03/21 2904<br>RELT LI           |

# Section 2: The data file consists of Blank Rows (total rows = 144836). Removing blank rows to keep the data.

```
bank_2 = bank_1.dropna(how='all')
bank_2.tail()
# Commenting Below link to avoid running it multiple time, please uncomment it once while rur
# del bank_2['month name']
bank_2.to_csv('/content/drive/MyDrive/Bank_Data/Bank_data_v2.csv', index=False)

bank_3 = pd.read_csv('/content/drive/MyDrive/Bank_Data/Bank_data_v2.csv')
bank_3.tail(5)
```

|              | Lead<br>ID | bankid | bank_account_id | account_number | Industry                                   | post_date | descrip <sup>†</sup>                        |
|--------------|------------|--------|-----------------|----------------|--|-----------|---|
| 29024        | 330698     | 8545   | 14374           | 1693           | Health<br>Care and<br>Social<br>Assistance | 06-Feb-17 | Wire trai<br>withdr<br>JAN<br>ASSOCIA       |
| 29025        | 330698     | 8545   | 14374           | 1693           | Health<br>Care and<br>Social<br>Assistance | 07-Feb-17 | ACH de<br>BANKC.<br>BTOT<br>PERSI<br>DOCTOF |
| Saved succes | sfully!    |        | ×               |                | Health                                     |           | Withdr                                      |

Section 3: Understand the Descriptive stats for the data.

Analyse and identify the patterns between variables.

Analyse the outliers and correlations between variables which might produce bias in the data.

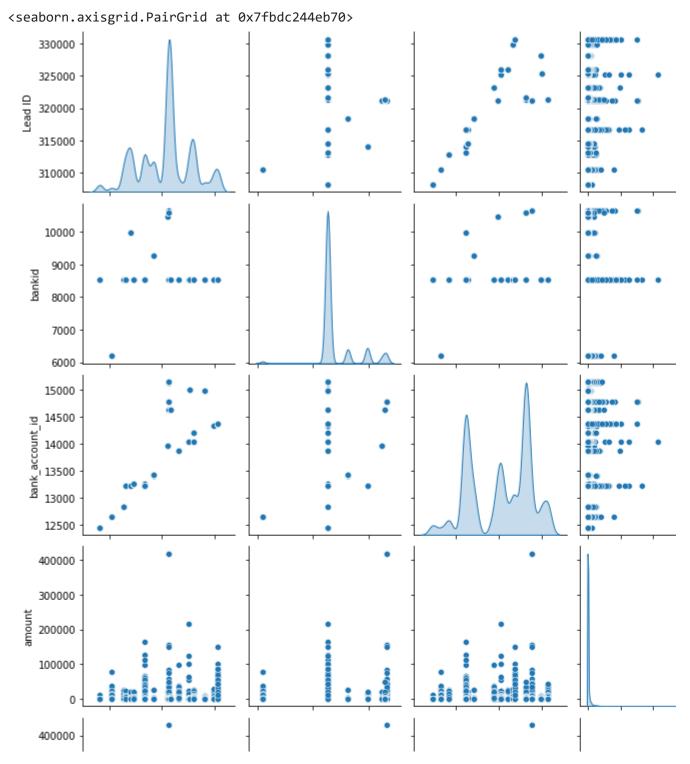
Double-click (or enter) to edit

del bank\_3['month name']
bank 3.describe()

|       | Lead ID       | bankid       | bank_account_id | amount        | running_balance | tr  |
|-------|---------------|--------------|-----------------|---------------|-----------------|-----|
| count | 29029.000000  | 29029.000000 | 29029.000000    | 29029.000000  | 29029.000000    | 290 |
| mean  | 320714.718282 | 8823.888491  | 14052.030556    | 1927.453475   | 10247.851975    |     |
| std   | 5051.317064   | 683.963276   | 703.164708      | 7450.638452   | 20055.683105    |     |
| min   | 308148.000000 | 6192.000000  | 12460.000000    | 0.000000      | -84727.980000   |     |
| 25%   | 316728.000000 | 8534.000000  | 13234.000000    | 29.250000     | 798.850000      |     |
| 50%   | 321380.000000 | 8535.000000  | 14049.000000    | 160.500000    | 3820.940000     |     |
| 75%   | 323253.000000 | 8545.000000  | 14636.000000    | 788.460000    | 11898.880000    |     |
| max   | 330698 000000 | 10656 000000 | 15148 000000    | 419000 000000 | 437942 290000   |     |

sns.pairplot(bank\_3, diag\_kind='kde')



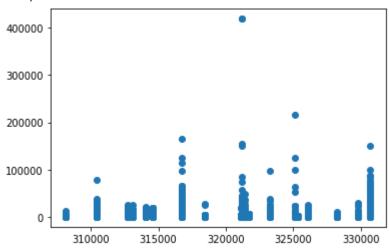


<matplotlib.axes.\_subplots.AxesSubplot at 0x7fbdcd676550>



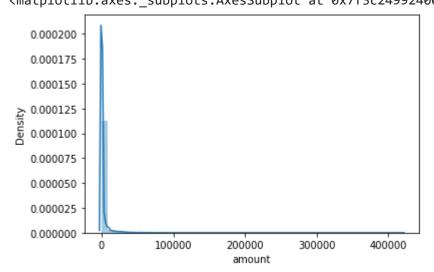
plt.scatter( bank\_3['Lead ID'],bank\_3['amount'])

<matplotlib.collections.PathCollection at 0x7fbdcc783e80>



sns.distplot(bank\_3['amount'])

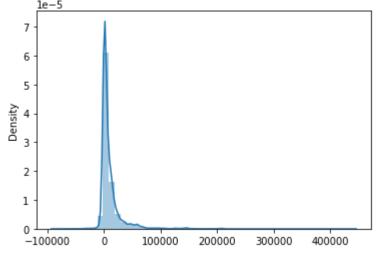
/usr/local/lib/python3.6/dist-packages/seaborn/distributions.py:2551: FutureWarning: `d
 warnings.warn(msg, FutureWarning)
<matplotlib.axes.\_subplots.AxesSubplot at 0x7f3c24992400>



sns.distplot(bank\_3['running\_balance'])

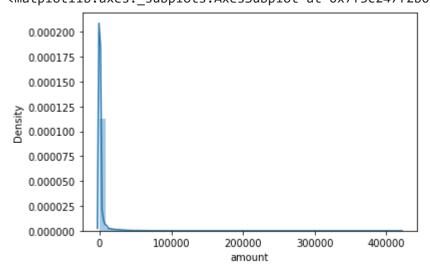
/usr/local/lib/python3.6/dist-packages/seaborn/distributions.py:2551: FutureWarning: `d warnings.warn(msg, FutureWarning)

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f3c2493b4e0>

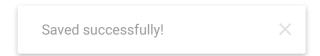


sns.distplot(bank\_3['amount'])

/usr/local/lib/python3.6/dist-packages/seaborn/distributions.py:2551: FutureWarning: `d
 warnings.warn(msg, FutureWarning)
<matplotlib.axes. subplots.AxesSubplot at 0x7f3c247f2b00>



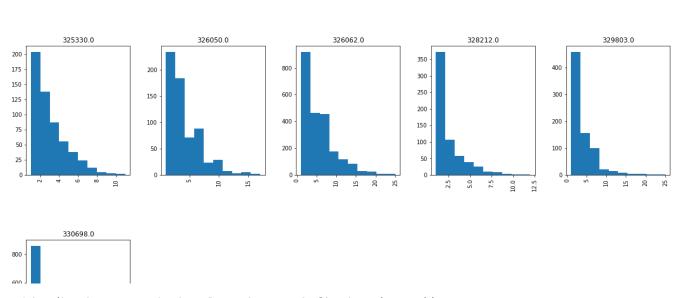
- Section 4: From the 2 histogram plots below, we understand the pattern between :
  - 1. Lead ID and Trans\_order Out of 21 Lead IDs, 4 of them have more than 30 transactions on their accounts.



### 2. Trans\_order Vs Amount - Majority of transaction amount is

bank\_3.hist(by='Lead ID',column='trans\_order',figsize=(20,30))

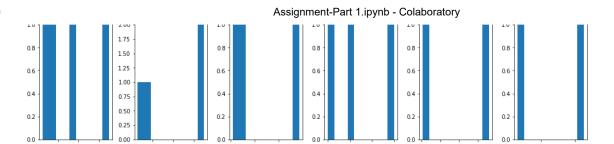
```
array([[<matplotlib.axes. subplots.AxesSubplot object at 0x7f3c24790e48>,
              <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c2466ffd0>,
              <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c2462d3c8>,
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             [<matplotlib.axes. subplots.AxesSubplot object at 0x7f3c245c1e48>,
             <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c24580208>,
             <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c24532550>,
             <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c245325c0>,
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             <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c242a7fd0>],
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              <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c24296748>,
              <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c24248ac8>,
             <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c241fbe48>,
              <matplotlib.axes._subplots.AxesSubplot object at 0x7f3c241ba208>]],
           dtype=object)
             308148.0
                                                     312745.0
                                 310443.0
                                                                          313082.0
                                                                                              314036.0
     250
                                                                  350
                                                                                       800
                         120
                                                                  300
                                                                                       700
     200
                         100
                                                                                       600
                                                                  250
                                              300
                          80
     150
                                                                                       500
                                                                  200
                                                                                       400
                          60
                                              200
     100
                                                                                       300
                          40
                                                                  100
                                                                                       200
     50
                                              100
                          20
                                                                   50
                                                                                       100
             314559 0
                                 316728.0
                                                     318465.0
                                                                          321146.0
                                                                                               321218.0
                                             1000
     160
                         1400
                                                                  300
     140
                                                                                       350
                         1200
     120
                                                                  250
                                                                                       300
                         1000
     100
                                              600
                                                                  200
                                                                                       250
                         800
     80
                                                                                       200
                                                                  150
                         600
                                              400
     60
                                                                                       150
                         400
                                                                  100
     40
                                                                                       100
                                              200
                                                                   50
                         200
     20
                                                                                       50
             321356.0
                                 321380.0
                                                     321671.0
                                                                          323253.0
                                                                                              325142.0
     600
                         2000
                                             1000
                                                                                       400
                                                                  250
                         1750
     500
                                              800
                         1500
                                                                  200
                                                                                       300
     400
                         1250
                                              600
                                                                                       250
                         1000
                                                                                       200
                                              400
                                                                  100
                                                                                       150
Saved successfully!
                                                                                       100
                                              200
                                                                   50
                                                                                       50
```



bank\_3.hist(by='trans\_order',column='amount',figsize=(20,40))
#From the 2 histogram plots, we understand the pattern between :

#1. Lead ID and Trans\_order - Out of 21 Lead IDs, 4 of them have more than 30 transactions or
#2. Trans\_order Vs Amount - Majority of transaction amount is till 21 transactions (all amour

```
array([[<matplotlib.axes. subplots.AxesSubplot object at 0x7f3c2400fdd8>,
            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c23c5f6a0>,
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            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c237cfac8>,
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            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c23738f98>],
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            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c236ab4a8>,
            <matplotlib.axes._subplots.AxesSubplot object at 0x7f3c23660710>,
            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c23694978>,
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            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c235be0f0>],
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            <matplotlib.axes._subplots.AxesSubplot object at 0x7f3c235235c0>,
            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c23557828>,
            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c2350ba90>,
            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c234bdcf8>,
            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c23474f60>,
            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c2342ad30>],
           [<matplotlib.axes._subplots.AxesSubplot object at 0x7f3c233eb0f0>,
            <matplotlib.axes._subplots.AxesSubplot object at 0x7f3c2339c470>,
            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c233ce7f0>,
            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c23381b70>,
            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c23332ef0>,
            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c232f22b0>,
            <matplotlib.axes. subplots.AxesSubplot object at 0x7f3c232a2630>]],
          dtype=object)
                                                                           6.0
                                                                                        7.0
    6000
                                                                                 1400
                              3500
                                           2500
                                                        2000
                 4000
                                                                                 1200
                              3000
    5000
                                                                     1500
                              2500
                                                                                 1000
                                           2000
                                                                     1250
                                                                                  800
                                                                     1000
                                           1500
                                                        1000
                                                                                  600
                                                                     750
Saved successfully!
                                           1000
                                                                                  400
                                                                     500
                                           500
                                                                                  200
```



```
# Compute the correlation matrix
corr = bank_3.corr()
print(corr)
```

```
Lead ID
                             bankid
                                          running_balance trans_order
Lead ID
                 1.000000 -0.158936
                                                  0.063881
                                                              -0.008326
bankid
                -0.158936
                          1.000000
                                                 -0.053088
                                                               0.001317
bank_account_id 0.700699 -0.050178
                                                 -0.041772
                                                               0.076003
amount
                 0.179994 -0.057082
                                                  0.181640
                                                              -0.078309
running_balance 0.063881 -0.053088
                                                  1.000000
                                                               0.022697
trans_order
                -0.008326
                          0.001317
                                                  0.022697
                                                               1.000000
```

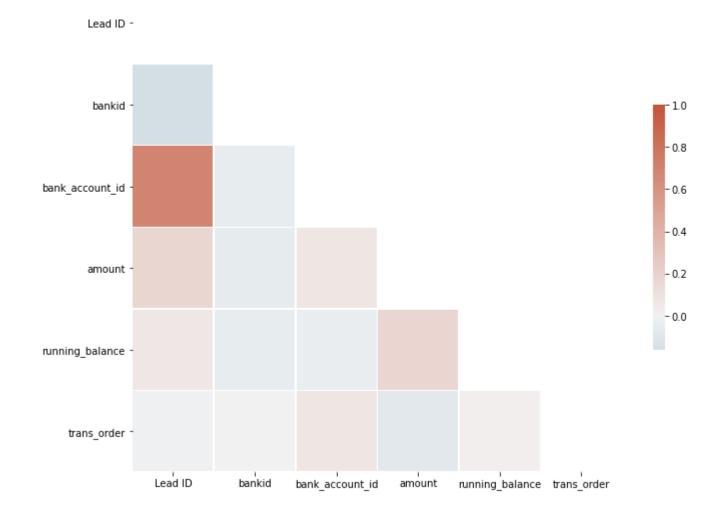
[6 rows x 6 columns]

```
# Generate a mask for the upper triangle
mask = np.triu(np.ones like(corr. dtype=bool))
Saved successfully!

f ax = nlt subplots(figsize=(11 9))
```

27/12/2020

<matplotlib.axes.\_subplots.AxesSubplot at 0x7fbdc0dfb0f0>



Data Summarization: To generate outputs as per requirement

Section 5: **Problem Statement**: This is a sample of individual bank statement transactions from 20 small business merchants (indexed by Lead ID) from various industries

Each merchant could have multiple accounts (bank account

Transactions may be debits (withdrawals) or credits (deposits)

Each transaction also has a description

→ 1(i) the number of bank accounts for each merchant

```
# 1. (i) the number of bank accounts for each merchant
Data_1 = bank_3.groupby('Lead ID')['bank_account_id'].nunique()
print(Data 1)
Data 1.to csv('/content/drive/MyDrive/Bank Data/Data 1.csv', index=True)
     Lead ID
     308148
     310443
               2
     312745
     313082
     314036
               1
     314559
     316728
     318465
               7
     321146
     321218
               1
     321356
               5
     321380
     321671
     323253
     325142
               1
     325330
               1
     326050
     326062
     328212
     329803
               1
     330698
     Name: bank_account_id, dtype: int64
```

1(ii) the number of months of each bank account for which data is available

```
Saved successfully! X ach bank account for which data is available
```

```
#bank_3['month_year'] = pd.to_datetime(bank_3['post_date']).dt.to_period('M')
Data_2 = bank_3.groupby('bank_account_id')['month_year'].nunique()
print(Data 2)
Data_2.to_csv('/content/drive/MyDrive/Bank_Data_2.csv', index=True)
     bank_account_id
     12460
               10
     12654
                7
     12655
               12
     12835
                5
                5
     12836
     12837
                1
     12838
               11
     12839
               13
     13226
               11
     13228
               12
                7
     13230
     13232
               13
     13233
               10
               13
     13234
     13235
               11
     13271
               12
     13272
                8
                7
     13273
                9
     13419
     13420
                3
                2
     13421
     13422
                7
                8
     13423
                9
     13424
                8
     13425
     13877
               11
     13879
                8
     13970
                9
     13971
                4
     14044
               11
     14045
                2
     14046
                8
     14049
                8
     14206
                8
     14339
               13
     14374
               13
     14629
               13
     14630
                2
     14631
                2
                2
     14632
                2
     14633
     14634
               12
               12
     14635
     14636
               13
               13
     14637
     14779
               13
     14981
               13
```

https://colab.research.google.com/drive/1ZtM7r3F7pKcDewTs20qjfVZhGReFeWi1#scrollTo=GBjZl5Su-nxA&printMode=true

15146 13 15147 11 15148 11

Name: month\_year, dtype: int64

### 1(iii) the total number of credits (deposits), debits

 (withdrawals) and their averages per month for each bank account and each merchant

```
# 1. (iii) the total number of credits (deposits), debits (withdrawals) and their averages per # Mean refers to the averages

Data_3 = pd.pivot_table(bank_3,index=['Lead ID','bank_account_id','month_year'],columns=['traprint(Data_3.round(decimals=0))

Data_3.to_csv('/content/drive/MyDrive/Bank_Data/Data_3.csv', index=True)
```

| transacti | on_type        |            | len<br>amount<br>credit | debit | mean<br>amount<br>credit | debit   |
|-----------|----------------|------------|-------------------------|-------|--------------------------|---------|
| Lead ID b | ank_account_id | month_year |                         |       |                          |         |
| 308148 1  | 2460           | 2016-03    | 5.0                     | 15.0  | 1220.0                   | 333.0   |
|           |                | 2016-04    | 1.0                     | 12.0  | 1500.0                   | 216.0   |
|           |                | 2016-05    | 4.0                     | 8.0   | 194.0                    | 99.0    |
|           |                | 2016-06    | 4.0                     | 13.0  | 592.0                    | 156.0   |
|           |                | 2016-07    | 20.0                    | 30.0  | 405.0                    | 280.0   |
| • • •     |                |            |                         |       | • • •                    | • • •   |
| 330698 1  | 4374           | 2016-10    | 27.0                    | 77.0  | 33553.0                  | 11766.0 |
|           |                | 2016-11    | 26.0                    | 85.0  | 32040.0                  | 9800.0  |
|           |                | 2016-12    | 33.0                    | 93.0  | 27355.0                  | 9495.0  |
|           |                | 2017-01    | 32.0                    | 196.0 | 28218.0                  | 4656.0  |
|           |                | 2017-02    | 6.0                     | 38.0  | 32059.0                  | 5011.0  |

[478 rows x 4 columns]

- 1(iv) the total dollar value of credits, debits and their
- averages per month for each bank account and each merchant

### Mean refers to the averages

print(Data\_4.round(decimals=2))
Data\_4.to\_csv('/content/drive/MyDrive/Bank\_Data\_4.csv', index=True)

|                        |              | sum       |           | mean     |          |
|------------------------|--------------|-----------|-----------|----------|----------|
|                        |              | amount    |           | amount   |          |
| transaction_type       |              | credit    | debit     | credit   | debit    |
| Lead ID bank_account_i | d month_year |           |           |          |          |
| 308148 12460           | 2016-03      | 6098.62   | 4995.32   | 1219.72  | 333.02   |
|                        | 2016-04      | 1500.00   | 2595.75   | 1500.00  | 216.31   |
|                        | 2016-05      | 774.95    | 791.74    | 193.74   | 98.97    |
|                        | 2016-06      | 2369.38   | 2028.87   | 592.34   | 156.07   |
|                        | 2016-07      | 8105.04   | 8396.89   | 405.25   | 279.90   |
| •••                    |              |           |           |          |          |
| 330698 14374           | 2016-10      | 905930.37 | 905998.55 | 33552.98 | 11766.21 |
|                        | 2016-11      | 833031.28 | 833031.28 | 32039.66 | 9800.37  |
|                        | 2016-12      | 902727.28 | 883050.45 | 27355.37 | 9495.17  |
|                        | 2017-01      | 902978.48 | 912591.12 | 28218.08 | 4656.08  |
|                        | 2017-02      | 192354.78 | 190403.36 | 32059.13 | 5010.61  |

[478 rows x 4 columns]

1. (v) Aggregate the answers to (iii) and (iv) at the merchant level, industry level and bankid level

# 1(v) Aggregate the answers to (iii) and (iv) at the merchant level, industry level and bankid level

#1. (v) Aggregate the answers to (iii) and (iv) at the merchant level, industry level and bar
Data\_5 = pd.pivot\_table(bank\_3,index=['Lead ID','Industry','bankid'],columns=['transaction\_ty
print(Data\_5.round(decimals=0))

Data\_5.to\_csv('/content/drive/MyDrive/Bank\_Data\_5.csv', index=True)

|   |        | len    | <br>mean   |
|---|--------|--------|------------|
|   |        | amount | <br>amount |
| transaction_type  |        | credit | <br>debit  |
| Lead ID Industry  | bankid |        |            |
| 308148 Accommodation and Food Services                  | 8535   | 172.0  | <br>193.0  |
| 310443 Construction                                     | 6192   | 35.0   | <br>1409.0 |
| 312745 Professional, Scientific, and Technical Services | 8544   | 160.0  | <br>1189.0 |
| 313082 Professional, Scientific, and Technical Services | 8535   | 112.0  | <br>545.0  |
| 314036 Retail Trade                                     | 9966   | 276.0  | <br>218.0  |
| 314559 Information Technology                           | 8534   | 143.0  | <br>1830.0 |
| 316728 Construction                                     | 8534   | 30.0   | <br>1350.0 |
|   | 8544   | 268.0  | <br>1031.0 |
| 318465 Health Care and Social Assistance                | 9262   | 369.0  | <br>136.0  |
| 321146 Retail Trade                                     | 10479  | 121.0  | <br>222.0  |
| 321218 Agriculture, Forestry, Fishing and Hunting       | 10656  | 57.0   | <br>1727.0 |
| 321356 Other Services (except Public Administration)    | 8544   | 697.0  | <br>5257.0 |
|   | 8534   | 4436.0 | <br>3661.0 |
| Saved successfully!                                     | 10591  | 503.0  | <br>1879.0 |
|   | 8534   | 642.0  | <br>172.0  |

[23 rows x 6 columns]

```
323253 Retail Trade
                                                         8545
                                                                  308.0 ...
                                                                               2330.0
325142 Other Services (except Public Administration)
                                                         8535
                                                                  115.0 ...
                                                                                529.0
325330 Retail Trade
                                                         8544
                                                                  312.0 ...
                                                                                312.0
326050 Accommodation and Food Services
                                                         8534
                                                                  190.0
                                                                                591.0
326062 Accommodation and Food Services
                                                         8535
                                                                  852.0 ...
                                                                                587.0
328212 Educational Services
                                                                  282.0 ...
                                                         8535
                                                                                507.0
329803 Construction
                                                         8535
                                                                  116.0
                                                                                324.0
330698 Health Care and Social Assistance
                                                         8545
                                                                  346.0 ... 12063.0
```

2 Consider Lead ID: 318465, bank\_account\_id: 13419 - plot the withdrawals, deposits and daily balance as a daily time series; do the same for Lead ID: 326062, bank\_account\_id: 14046

```
# 2. Consider Lead ID: 318465, bank account id: 13419 - plot the withdrawals, deposits and c
Data 6 = bank 3[(bank 3['Lead ID'] == 318465) & (bank 3.bank account id == 13419)]
Data_6a = Data_6[(bank_3['transaction_type'] == 'debit')]
Data_6c = Data_6[(bank_3['transaction_type'] == 'credit')]
Data_6b = pd.pivot_table(Data_6a,columns=['post_date'],values=['amount'],aggfunc=[np.sum])
print(Data 6a, Data 6b)
Data 6.to csv('/content/drive/MyDrive/Bank Data/Data 6.csv', index=True)
#pd.pivot_table(bank_3,index=['Lead ID','Industry','bankid'],columns=['transaction_type'],val
#Plot as daily time series for daily withdrawals.
import matplotlib.pyplot as plt
from pylab import rcParams
rcParams['figure.figsize'] = 100,30
plt.plot(Data 6a['post date'], Data 6a['amount'],color='red', marker='o')
plt.title('Daily withdrawal Time series', fontsize=8)
plt.xlabel('Date', fontsize=14)
plt.ylabel('Daily withdrawal', fontsize=8)
plt.show()
#Plot as daily time series for deposit.
rcParams['figure.figsize'] = 100,30
plt.plot(Data_6c['post_date'], Data_6c['amount'],color='red', marker='o')
plt.title('Daily Deposit Time series', fontsize=8)
 Saved successfully!
                                X ==8)
```

```
#Plot as daily time series for net balance.
rcParams['figure.figsize'] = 100,30
plt.plot(Data 6a['post date'], Data 6a['amount'],color='red', marker='o')
plt.title('Daily Time series', fontsize=8)
plt.xlabel('Date', fontsize=8)
plt.ylabel('Daily Balance', fontsize=14)
plt.show()
                                               Traceback (most recent call last)
     NameError
     <ipython-input-4-b52411c48272> in <module>()
           1 # 2. Consider Lead ID: 318465, bank account id: 13419 - plot the withdrawals,
     deposits and daily balance as a daily time series; do the same for Lead ID: 326062,
     bank account id: 14046
     ---> 3 Data 6 = bank 3[(bank 3['Lead ID'] == 318465) & (bank 3.bank account id ==
     13419)]
           4 Data 6a = Data 6[(bank 3['transaction type'] == 'debit')]
           5 Data 6c = Data 6[(bank 3['transaction type'] == 'credit')]
     NameError: name 'bank_3' is not defined
# 2. Consider Lead ID: 326062, bank_account_id: 14046 - plot the withdrawals, deposits and c
Data 61 = bank 3[(bank 3['Lead ID'] == 326062) & (bank 3.bank account id == 14046)]
Data_61a = Data_61[(bank_3['transaction_type'] == 'debit')]
Data 61c = Data 61[(bank 3['transaction type'] == 'credit')]
Data 61b = pd.pivot table(Data 6a,columns=['post date'],values=['amount'],aggfunc=[np.sum])
print(Data 61)
Data 61.to csv('/content/drive/MyDrive/Bank Data/Data 61.csv', index=True)
#pd.pivot_table(bank_3,index=['Lead ID','Industry','bankid'],columns=['transaction_type'],val
#Plot as daily time series for daily withdrawals.
import matplotlib.pyplot as plt
from pylab import rcParams
rcParams['figure.figsize'] = 100,30
plt.plot(Data 61a['post date'], Data 61a['amount'],color='red', marker='o')
plt.title('Daily withdrawal Time series', fontsize=8)
plt.xlabel('Date', fontsize=14)
plt.ylabel('Daily withdrawal', fontsize=14)
plt.show()
#Plot as daily time series for deposit.
rcParams['figure.figsize'] = 100,30
plt.plot(Data_61c['post_date'], Data_61c['amount'],color='red', marker='o')
plt.title('Daily Deposit Time series', fontsize=8)
                                × ≥=14)
 Saved successfully!
```

```
#Plot as daily time series for net balance.

rcParams['figure.figsize'] = 100,30

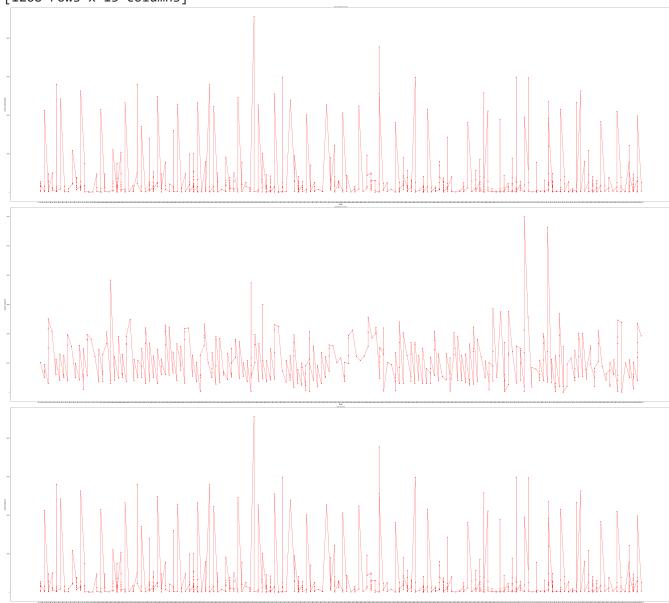
plt.plot(Data_61a['post_date'], Data_61a['amount'],color='red', marker='o')
plt.title('Daily Time series', fontsize=8)
plt.xlabel('Date', fontsize=14)
plt.ylabel('Daily Balance', fontsize=14)
plt.show()
```

/usr/local/lib/python3.6/dist-packages/ipykernel\_launcher.py:4: UserWarning: Boolean Se after removing the cwd from sys.path.

/usr/local/lib/python3.6/dist-packages/ipykernel\_launcher.py:5: UserWarning: Boolean Se

|       | Lead ID | bankid | <br>Primary_key                | month_year |
|-------|---------|--------|--------------------------------|------------|
| 25111 | 326062  | 8535   | <br>326062_8535_14046_42559_1  | 2016-07    |
| 25112 | 326062  | 8535   | <br>326062_8535_14046_42562_10 | 2016-07    |
| 25113 | 326062  | 8535   | <br>326062_8535_14046_42562_7  | 2016-07    |
| 25114 | 326062  | 8535   | <br>326062_8535_14046_42562_9  | 2016-07    |
| 25115 | 326062  | 8535   | <br>326062_8535_14046_42562_4  | 2016-07    |
|       |         |        | <br>•••                        |            |
| 26374 | 326062  | 8535   | <br>326062_8535_14046_42788_13 | 2017-02    |
| 26375 | 326062  | 8535   | <br>326062_8535_14046_42789_1  | 2017-02    |
| 26376 | 326062  | 8535   | <br>326062_8535_14046_42789_3  | 2017-02    |
| 26377 | 326062  | 8535   | <br>326062_8535_14046_42789_2  | 2017-02    |
| 26378 | 326062  | 8535   | <br>326062_8535_14046_42789_4  | 2017-02    |

#### [1268 rows x 13 columns]



- 3 For the merchants with the above Lead IDs, plot the
- withdrawals, deposits and daily balance as a daily time series (aggregate over all their bank accounts)

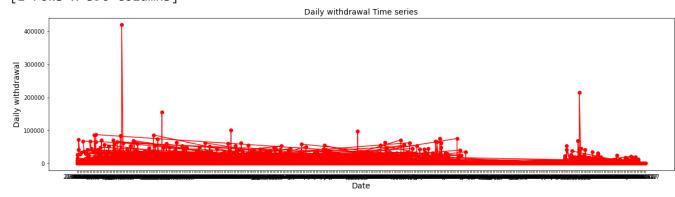
```
Data_7 = bank_3
Data_7a = Data_7[(bank_3['transaction_type'] == 'debit')]
Data_7c = Data_7[(bank_3['transaction_type'] == 'credit')]
Data 7b = pd.pivot table(Data 7a,columns=['post date'],values=['amount'],aggfunc=[np.sum])
print(Data_7a, Data_7b)
Data 7a.to csv('/content/drive/MyDrive/Bank Data/Data 7a.csv', index=True)
Data_7b.to_csv('/content/drive/MyDrive/Bank_Data/Data_7b.csv', index=True)
Data_7c.to_csv('/content/drive/MyDrive/Bank_Data/Data_7c.csv', index=True)
#pd.pivot table(bank 3,index=['Lead ID','Industry','bankid'],columns=['transaction type'],val
#Plot as daily time series for daily withdrawals.
import matplotlib.pyplot as plt
from pylab import rcParams
rcParams['figure.figsize'] = 20,5
plt.plot(Data_7a['post_date'], Data_7a['amount'],color='red', marker='o')
plt.title('Daily withdrawal Time series', fontsize=14)
plt.xlabel('Date', fontsize=14)
plt.ylabel('Daily withdrawal', fontsize=14)
plt.show()
#Plot as daily time series for deposit.
plt.plot(Data_7c['post_date'], Data_7c['amount'],color='red', marker='o')
plt.title('Daily Deposit Time series', fontsize=14)
plt.xlabel('Date', fontsize=14)
plt.ylabel('Daily Deposit', fontsize=14)
plt.show()
#Plot as daily time series for net balance.
rcParams['figure.figsize'] = 20,5
plt.plot(Data_7a['post_date'], Data_7a['amount'],color='red', marker='o')
plt.title('Daily Time series', fontsize=14)
plt.xlabel('Date', fontsize=14)
                                   ≥=14)
 Saved successfully!
```

```
Primary_key month_year
       Lead ID
                bankid
3
        308148
                   8535
                                308148_8535_12460_42450_2
                                                              2016-03
4
        308148
                   8535
                                308148_8535_12460_42451_1
                                                               2016-03
                         . . .
5
                   8535
                                308148 8535 12460 42452 1
                                                               2016-03
        308148
6
        308148
                   8535
                               308148 8535 12460 42453 3
                                                              2016-03
                   8535
7
                               308148_8535_12460_42453_1
        308148
                                                              2016-03
                    . . .
29023
        330698
                   8545
                               330698_8545_14374_42772_14
                                                              2017-02
29024
        330698
                   8545
                               330698 8545 14374 42772 7
                                                               2017-02
29026
                   8545
                               330698 8545 14374 42773 4
                                                              2017-02
        330698
29027
        330698
                   8545
                               330698 8545 14374 42773 3
                                                              2017-02
29028
        330698
                   8545
                               330698_8545_14374_42773_1
                                                              2017-02
[18487 rows x 13 columns]
                                            sum
post_date 01-Apr-16 01-Aug-16 01-Dec-15
                                             ... 31-Mar-16
                                                            31-May-16
                                                                        31-0ct-16
           67748.53
                     173552.05
```

8366.5

[1 rows x 396 columns]

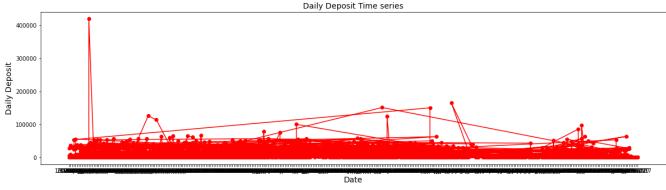
amount

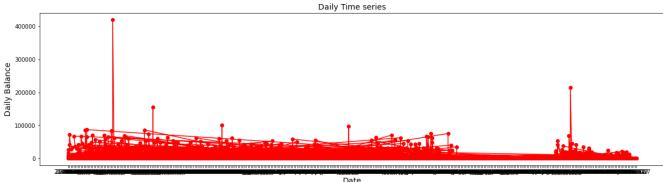


97632.7

210444.38

148666.72





#### 4 Top 5 and Bottom 5 Merchants/Banks

```
#Top 5 Merchant IDs by amount
#del bank 3['month name']
#bank 3.to csv('/content/drive/MyDrive/Bank Data/Bank Data v3.csv', index=False)
bank_4 = bank_3.sort_values('running_balance',ascending = False).groupby('Lead ID').head(2)
print(bank_4.head(5))
bank 4.to csv('/content/drive/MyDrive/Bank Data/top5 lead id.csv', index=False)
#Bottom 5 Merchant IDs by amount
bank_4 = bank_3.sort_values('running_balance',ascending = True).groupby('Lead ID').head(2)
print(bank 4.head(5))
bank 4.to csv('/content/drive/MyDrive/Bank Data/Bottom5 lead id.csv', index=False)
#Top 5 bank IDs by amount
bank_4 = bank_3.sort_values('running_balance',ascending = False).groupby('bankid').head(2)
print(bank 4.head(5))
bank_4.to_csv('/content/drive/MyDrive/Bank_Data/top5_bankid.csv', index=False)
#Bottom 5 bank IDs by amount
bank_4 = bank_3.sort_values('running_balance',ascending = True).groupby('bankid').head(2)
print(bank 4.head(5))
bank 4.to csv('/content/drive/MyDrive/Bank Data/bottom5 bankid.csv', index=False)
 Saved successfully!
                                                 Primary_key month_year
                                   21218 10656 14779 42552 1
                                                                2016-07
```