## Import Library

```
1 !pip install gdown
2 !pip install matplotlib
3 !pip install wget
4 import gdown
5 import pandas as pd
6 import numpy as np
7 import os
8 import math
9 import scipy.stats as stat
10 import datetime as dt
11 import seaborn as sb
12 import re
13 import matplotlib
14 import matplotlib.pyplot as plt
15 %matplotlib inline
    Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/public/simple/</a>
    Requirement already satisfied: gdown in /usr/local/lib/python3.8/dist-packages (4.4.0)
    Requirement already satisfied: requests[socks] in /usr/local/lib/python3.8/dist-packages (from gdown) (2.25.1)
    Requirement already satisfied: beautifulsoup4 in /usr/local/lib/python3.8/dist-packages (from gdown) (4.6.3)
    Requirement already satisfied: six in /usr/local/lib/python3.8/dist-packages (from gdown) (1.15.0)
    Requirement already satisfied: tqdm in /usr/local/lib/python3.8/dist-packages (from gdown) (4.64.1)
    Requirement already satisfied: filelock in /usr/local/lib/python3.8/dist-packages (from gdown) (3.9.0)
    Requirement already satisfied: urllib3<1.27,>=1.21.1 in /usr/local/lib/python3.8/dist-packages (from requests[socks]->gdown) (1.24
    Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.8/dist-packages (from requests[socks]->gdown) (2.10)
    Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.8/dist-packages (from requests[socks]->gdown) (2022.12
    Requirement already satisfied: chardet<5,>=3.0.2 in /usr/local/lib/python3.8/dist-packages (from requests[socks]->gdown) (4.0.0)
    Requirement already satisfied: PySocks!=1.5.7,>=1.5.6 in /usr/local/lib/python3.8/dist-packages (from requests[socks]->gdown) (1.7
    Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/public/simple/</a>/
    Requirement already satisfied: matplotlib in /usr/local/lib/python3.8/dist-packages (3.2.2)
    Requirement already satisfied: python-dateutil>=2.1 in /usr/local/lib/python3.8/dist-packages (from matplotlib) (2.8.2)
    Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.8/dist-packages (from matplotlib) (1.4.4)
    Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.8/dist-packages (from matplotlib) (0.11.0)
     Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in /usr/local/lib/python3.8/dist-packages (from matplotlib)
    Requirement already satisfied: numpy>=1.11 in /usr/local/lib/python3.8/dist-packages (from matplotlib) (1.21.6)
    Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.8/dist-packages (from python-dateutil>=2.1->matplotlib) (1.15.0)
    Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/public/simple/</a>
    Requirement already satisfied: wget in /usr/local/lib/python3.8/dist-packages (3.2)
```

## Import Data

```
1 url1 = "https://raw.githubusercontent.com/hendraswntr/MiniProject-DataAnalyst-RFM_Analysis/main/data/Products.csv"
2 url2 = "https://raw.githubusercontent.com/hendraswntr/MiniProject-DataAnalyst-RFM_Analysis/main/data/OrderDetails.csv"
3 url3 = "https://raw.githubusercontent.com/hendraswntr/MiniProject-DataAnalyst-RFM_Analysis/main/data/Orders.csv"
4 url4 = "https://raw.githubusercontent.com/hendraswntr/MiniProject-DataAnalyst-RFM_Analysis/main/data/Customers.csv"
5
6 products = pd.read_csv(url1, sep=';')
7 orderdetails = pd.read_csv(url2, sep=';')
8 orders = pd.read_csv(url3, sep=';')
9 customers = pd.read_csv(url4, sep=';')
10
```

1 products.head()

	ProductID	ProductName	SupplierID	CategoryID	QuantityPerUnit	UnitPrice	Uni
0	1	Chai	1	1	10 boxes x 20 bags	18	
1	2	Chang	1	1	24 - 12 oz bottles	19	
2	3	Aniseed Syrup	1	2	12 - 550 ml bottles	10	
3	4	Chef Anton's Cajun Seasoning	2	2	48 - 6 oz jars	22	
4	5	Chef Anton's Gumbo Mix	2	2	36 boxes	21,35	

1 orderdetails.head()

	OrderID	ProductID	UnitPrice	Quantity	Discount	7
0	10248	11	14	12	0	
1	10248	42	9,8	10	0	
2	10248	72	34,8	5	0	
3	10249	14	18,6	9	0	
4	10249	51	42,4	40	0	

1 orders.head()

	OrderID	CustomerID	EmployeeID	OrderDate	RequiredDate	ShippedDate	ShipVia
0	10248	VINET	5	1996-07- 04 00:00:00	1996-08-01 00:00:00	1996-07-16 00:00:00	3
1	10249	TOMSP	6	1996-07- 05 00:00:00	1996-08-16 00:00:00	1996-07-10 00:00:00	1
2	10250	HANAR	4	1996-07- 08 00:00:00	1996-08-05 00:00:00	1996-07-12 00:00:00	2
3	10251	VICTE	3	1996-07- 08 00:00:00	1996-08-05 00:00:00	1996-07-15 00:00:00	1
4	10252	SUPRD	4	1996-07- 09 00:00:00	1996-08-06 00:00:00	1996-07-11 00:00:00	2
7							
4							<b>&gt;</b>

1 customers.head()

	CustomerID	CompanyName	ContactName	ContactTitle	Address	City	Regio
0	ALFKI	Alfreds Futterkiste	Maria Anders	Sales Representative	Obere Str. 57	Berlin	Na
1	ANATR	Ana Trujillo Emparedados y helados	Ana Trujillo	Owner	Avda. de la Constitucion 2222	Moxico D.F.	Na
2	ANTON	Antonio Moreno Taqueroa	Antonio Moreno	Owner	Mataderos 2312	Moxico D.F.	Na
3	AROUT	Around the Horn	Thomas Hardy	Sales Representative	120 Hanover Sq.	London	Na
4	BERGS	Berglunds snabbkop	Christina Berglund	Order Administrator	Berguvsvogen 8	Luleo	Na
7							
4							-

# → Data Merge

```
1 data1 = pd.merge(customers, orders, on=['CustomerID'])
2 data2 = pd.merge(data1, orderdetails, on=['OrderID'])
3 data = pd.merge(data2, products, on=['ProductID'])
```

1 data.head()

	CustomerID	CompanyName	ContactName	ContactTitle	Address	City	Reg
0	ALFKI	Alfreds Futterkiste	Maria Anders	Sales Representative	Obere Str. 57	Berlin	ı
1	ALFKI	Alfreds Futterkiste	Maria Anders	Sales Representative	Obere Str. 57	Berlin	1
2	BERGS	Berglunds snabbkop	Christina Berglund	Order Administrator	Berguvsvogen 8	Luleo	1
3	BERGS	Berglunds snabbkop	Christina Berglund	Order Administrator	Berguvsvogen 8	Luleo	I

1 data.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 2155 entries, 0 to 2154
Data columns (total 37 columns):

Data	columns (total 3	37 columns):	
#	Column	Non-Null Count	Dtype
	 CtTD	245511	
0	CustomerID	2155 non-null	object
1	CompanyName	2155 non-null	object
2	ContactName	2155 non-null	object
3	ContactTitle	2155 non-null	object
4	Address	2155 non-null	object
5	City	2155 non-null	object
6	Region	826 non-null	object
7	PostalCode	2100 non-null	object
8	Country	2155 non-null	object
9	Phone	2155 non-null	object
10	Fax	1506 non-null	object
11	OrderID	2155 non-null	int64
12	EmployeeID	2155 non-null	int64
13	OrderDate	2155 non-null	object
14	RequiredDate	2155 non-null	object
15	ShippedDate	2082 non-null	object
16	ShipVia	2155 non-null	int64
17	Freight	2155 non-null	object
18	ShipName	2155 non-null	object
19	ShipAddress	2155 non-null	object
20	ShipCity	2155 non-null	object
21	ShipRegion	856 non-null	object
22	ShipPostalCode	2100 non-null	object
23	ShipCountry	2155 non-null	object
24	ProductID	2155 non-null	int64
25	UnitPrice_x	2155 non-null	object
26	Quantity	2155 non-null	int64
27	Discount	2155 non-null	object
28	ProductName	2155 non-null	object
29	SupplierID	2155 non-null	int64
30	CategoryID	2155 non-null	int64
31	QuantityPerUnit	2155 non-null	object
32	UnitPrice y	2155 non-null	object
33	UnitsInStock	2155 non-null	int64
34	UnitsOnOrder	2155 non-null	int64
35	ReorderLevel	2155 non-null	int64
36	Discontinued	2155 non-null	bool
dtype	es: bool(1), inte		)
	ry usage: 625.0+	. ,, ,	•
	, ,		

# ▼ Data Cleansing & Processing

1 data.isnull().sum()

CustomerID	0
CompanyName	0
ContactName	0
ContactTitle	0
Address	0
City	0
Region	1329
PostalCode	55
Country	0
Phone	0
Fax	649
OrderID	0
EmployeeID	0
OrderDate	0
RequiredDate	0
ShippedDate	73
ShipVia	0
Freight	0
ShipName	0
ShipAddress	0

```
ShipCity
                          0
                       1299
   ShipRegion
    ShipPostalCode
                         55
    ShipCountry
   ProductID
   UnitPrice_x
                          0
   Quantity
                          a
   Discount
                          0
   {\tt ProductName}
                          0
   SupplierID
                          0
   CategoryID
   QuantityPerUnit
                          0
   UnitPrice_y
                          0
   UnitsInStock
   UnitsOnOrder
   ReorderLevel
                          0
   Discontinued
   dtype: int64
1 del data['Region']
2 del data['PostalCode']
3 del data['Fax']
4 del data['ShippedDate']
5 del data['ShipRegion']
6 del data['ShipPostalCode']
1 data.duplicated().sum()
   0
```

### 1 data.nunique()

CustomerID CompanyName 89 ContactName 89 ContactTitle 12 Address 89 City 69 Country 21 Phone 89 OrderID 830 EmployeeID OrderDate 480 RequiredDate 454 ShipVia 3 Freight 799  ${\tt ShipName}$ 90 ShipAddress 89 ShipCity 70 ShipCountry 21 ProductID 77 UnitPrice\_x 116 Quantity 55 Discount 11 ProductName 77 SupplierID 29 CategoryID 8 QuantityPerUnit 70 UnitPrice\_y 62 UnitsInStock UnitsOnOrder 10 ReorderLevel Discontinued dtype: int64

#### 1 data.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 2155 entries, 0 to 2154
Data columns (total 31 columns):

Data	corumns (cocar 3	or corumns):	
#	Column	Non-Null Count	Dtype
0	CustomerID	2155 non-null	object
1	CompanyName	2155 non-null	object
2	ContactName	2155 non-null	object
3	ContactTitle	2155 non-null	object
4	Address	2155 non-null	object
5	City	2155 non-null	object
6	Country	2155 non-null	object
7	Phone	2155 non-null	object
8	OrderID	2155 non-null	int64
9	EmployeeID	2155 non-null	int64
10	OrderDate	2155 non-null	object
11	RequiredDate	2155 non-null	object
12	ShipVia	2155 non-null	int64

```
13
        Freight
                        2155 non-null
                                        object
    14 ShipName
                        2155 non-null
    15
       ShipAddress
                        2155 non-null
                                        object
    16 ShipCity
                       2155 non-null
                                        object
    17 ShipCountry
                        2155 non-null
                                        object
    18 ProductID
                        2155 non-null
                                        int64
    19 UnitPrice_x
                        2155 non-null
                                        object
                        2155 non-null
    20 Quantity
                                        int64
    21 Discount
                        2155 non-null
                                        object
    22 ProductName
                        2155 non-null
                                        object
    23 SupplierID
                        2155 non-null
                                        int64
    24
       CategoryID
                        2155 non-null
    25
       QuantityPerUnit 2155 non-null
                                        object
    26
       UnitPrice_y
                        2155 non-null
                                        object
       UnitsInStock
                        2155 non-null
                                        int64
    28 UnitsOnOrder
                        2155 non-null
                                        int64
    29 ReorderLevel
                        2155 non-null
                                        int64
    30 Discontinued
                        2155 non-null
                                        bool
   dtypes: bool(1), int64(10), object(20)
   memory usage: 524.0+ KB
1 \; data['UnitPrice_x'] = pd.to_numeric(data['UnitPrice_x'].str.replace(',',','), \; errors='coerce', \; downcast='float')
2 data['UnitPrice_y'] = pd.to_numeric(data['UnitPrice_y'].str.replace(',','.'), errors='coerce', downcast='float')
3 data['Discount'] = pd.to_numeric(data['Discount'].str.replace(',','.'), errors='coerce', downcast='float')
```

## Data Analysis

```
1 data['OrderDate'].head()
        1997-08-25 00:00:00
        1998-03-16 00:00:00
        1997-11-07 00:00:00
        1998-03-04 00:00:00
        1997-04-17 00:00:00
   Name: OrderDate, dtype: object
1 data['recentpurchasedate'] = pd.DatetimeIndex(data['OrderDate']).date
1 data['recentpurchasedate'].head()
        1997-08-25
        1998-03-16
        1997-11-07
        1998-03-04
        1997-04-17
   Name: recentpurchasedate, dtype: object
1 recency_data = data.groupby(by='CustomerID', as_index=False)['recentpurchasedate'].max()
2 recency_data.columns = ['CustomerID','lastpurshacedate']
3 recency_data.head()
       CustomerID lastpurshacedate
            ALFKI
                          1998-04-09
    0
           ANATR
                          1998-03-04
           ANTON
    2
                          1998-01-28
    3
           AROUT
                          1998-04-10
           BERGS
                          1998-03-04
1 data['OrderDate'].max()
   1998-05-06 00:00:00
1 now = dt.date(1998,5,6)
2 print(now)
   1998-05-06
1 recency_data['Recency'] = recency_data['lastpurshacedate'].apply(lambda x: (now - x).days)
2 recency_data.head()
```

```
        CustomerID
        lastpurshacedate
        Recency

        0
        ALFKI
        1998-04-09
        27

        1
        ANATR
        1998-03-04
        63

        2
        ANTON
        1998-01-28
        98
```

1 recency\_data.drop('lastpurshacedate',axis=1,inplace=True)

2 recency\_data.head()

	CustomerID	Recency	1
0	ALFKI	27	
1	ANATR	63	
2	ANTON	98	
3	AROUT	26	
4	BERGS	63	

- frequency\_data = data.groupby(by=['CustomerID'], as\_index=False)['OrderID'].count()
- 2 frequency\_data.columns = ['CustomerID','Frequency']
- 3 frequency\_data.head()

	CustomerID	Frequency	1
0	ALFKI	12	
1	ANATR	10	
2	ANTON	17	
3	AROUT	30	
4	BERGS	52	

```
1 data['TotalSales'] = (data['UnitPrice_x'] - data['Discount'])*data['Quantity']
2 data['TotalSales']
```

```
0 680.249977
1 91.199997
2 911.999969
3 1365.000000
4 136.799995
...
2150 185.399998
2151 386.249995
```

2151 386.249995 2152 155.000000 2153 247.999992 2154 775.000000

Name: TotalSales, Length: 2155, dtype: float64

```
1 monetary_data = data.groupby(by='CustomerID',as_index=False).agg({'TotalSales': 'sum'})
```

2 monetary\_data.columns = ['CustomerID','Monetary']

3 monetary\_data.head()

	CustomerID	Monetary
0	ALFKI	4583.500010
1	ANATR	1402.949990
2	ANTON	7497.149937
3	AROUT	13788.899798
4	BERGS	26900.049850

- 1 temp\_data = recency\_data.merge(frequency\_data,on='CustomerID')
- 2 rfm\_data = temp\_data.merge(monetary\_data,on='CustomerID')
- 3 rfm\_data.set\_index('CustomerID',inplace=True)
- 4 rfm data.head()

```
Monetary
                  Recency Frequency
      CustomerID
 1 rfm_data.shape
     (89, 3)
 1 quantiles = rfm_data.quantile(q=[0.25,0.5,0.75])
 2 quantiles
                                   Monetary
           Recency Frequency
      0.25
                          11.0
                                 3361.000004
                8.0
      0.50
               23.0
                          20.0
                                 7531.800077
                          31.0 18086.900101
      0.75
               58.0
 1 quantiles.to_dict()
     {'Recency': {0.25: 8.0, 0.5: 23.0, 0.75: 58.0},
      'Frequency': {0.25: 11.0, 0.5: 20.0, 0.75: 31.0},
      'Monetary': {0.25: 3361.0000038146973,
       0.5: 7531.8000774383545,
       0.75: 18086.900100708008}}
 1 def RScore(x,p,d):
       if x <= d[p][0.25]:
 3
          return 4
       elif x \leftarrow d[p][0.50]:
 4
 5
          return 3
 6
       elif x \leftarrow d[p][0.75]:
 7
          return 2
 8
       else:
 9
           return 1
10
11 def FMScore(x,p,d):
12
       if x <= d[p][0.25]:
13
           return 1
       elif x \leftarrow d[p][0.50]:
14
15
          return 2
       elif x <= d[p][0.75]:
16
17
          return 3
18
       else:
19
           return 4
 1 rfm_segmentation = rfm_data
 2 rfm_segmentation['R_Score'] = rfm_segmentation['Recency'].apply(RScore, args=('Recency',quantiles,))
 3 rfm_segmentation['F_Score'] = rfm_segmentation['Frequency'].apply(FMScore, args=('Frequency',quantiles,))
 4 rfm_segmentation['M_Score'] = rfm_segmentation['Monetary'].apply(FMScore, args=('Monetary',quantiles,))
 5 rfm_segmentation.head()
                                                                                 1
                  Recency Frequency
                                          Monetary R_Score F_Score M_Score
      CustomerID
        ALFKI
                                       4583.500010
                                                                             2
                       27
                                  12
        ANATR
                       63
                                  10
                                       1402.949990
                                                                    1
                                                                             1
       ANTON
                       98
                                       7497.149937
                                                                    2
                                                                             2
                                  17
       AROUT
                       26
                                      13788.899798
                                                                    3
                                                                             3
       BERGS
                                  52 26900.049850
                       63
                                                           1
                                                                    4
                                                                             4
 1 rfm_segmentation['RFMScore'] = rfm_segmentation.R_Score.map(str) \
 2
                               + rfm_segmentation.F_Score.map(str) \
                                + rfm_segmentation.M_Score.map(str)
 3
 4 rfm_segmentation.head()
```

CustomerID

```
Recency Frequency Monetary R_Score F_Score M_Score RFMScore
```

1 rfm\_segmentation[rfm\_segmentation['RFMScore']=='444'].sort\_values('Monetary', ascending=False).head(10)

	Recency	Frequency	Monetary	R_Score	F_Score	M_Score	RFMScore
CustomerID							
SAVEA	5	116	115253.339402	4	4	4	444
ERNSH	1	102	112903.829578	4	4	4	444
HUNGO	6	55	57109.540001	4	4	4	444
RATTC	0	71	52188.890416	4	4	4	444
QUEEN	2	40	30112.550216	4	4	4	444
WHITC	5	40	28990.799867	4	4	4	444
BONAP	0	44	23776.400169	4	4	4	444
HILAA	8	45	23577.579861	4	4	4	444
<b>LEHMS</b>	1	39	21209.819996	4	4	4	444

```
rfm_segmentation['Customer_Segmentation'] = np.where(rfm_segmentation['RFMScore'] ==444, "Best Customers",
                                      (np.where(rfm_segmentation['F_Score'] == 4,"Loyal Customers",
 2
 3
                                      (np.where(rfm_segmentation['M_Score'] == 4, "Big Spenders",
                                      (np.where(rfm_segmentation['F_Score'] == 3, "Gold Customers",
 4
 5
                                      (np.where(rfm_segmentation['F_Score'] == 2, "Silver Customers",
 6
                                      (np.where(rfm_segmentation['RFMScore'] == 244,"Almost Lost",
                                     (np.where(rfm_segmentation['RFMScore'] == 144, "Lost Customers",
 7
 8
                                     np.where(rfm_segmentation['RFMScore'] == 111,'Lost Cheap Customers', 'Low Value Customers')
9
                                              )))))))))))))))
10
     {\tt rfm\_segmentation}
```

```
Monetary R_Score F_Score M_Score RFMScore
             Recency Frequency
 CustomerID
   ALFKI
                  27
                             12
                                  4583.500010
                                                      2
                                                               2
                                                                        2
                                                                                222
   ANATR
                  63
                             10
                                  1402.949990
                                                                        1
                                                                                111
                                                               1
  ANTON
                                                                        2
                  98
                             17
                                  7497.149937
                                                      1
                                                               2
                                                                                122
  AROUT
                                 13788.899798
                                                      2
                                                                        3
                  26
                                                               3
                                                                                233
                             30
  BERGS
                  63
                                 26900.049850
                                                                        4
                                                                                144
  WARTH
                  21
                             37
                                 16580.599975
                                                      3
                                                                        3
                                                                                343
                                                               4
   WELLI
                                                                        2
                  58
                             19
                                  6465.750010
                                                      2
                                                               2
                                                                                222
   WHITC
                   5
                                                                        4
                             40 28990.799867
                                                      4
                                                               4
                                                                                444
   WILMK
                  29
                             17
                                  3161.349983
                                                      2
                                                               2
                                                                        1
                                                                                221
  WOLZA
                                                                        2
                  13
                             16
                                  3531.949996
                                                      3
                                                               2
                                                                                322
89 rows × 8 columns
```

```
1 rfsegment = pd.merge(rfm_segmentation, data, on=['CustomerID'])
2 data = rfsegment.sort_values(by=['CustomerID'])
3 data.info()
```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 2155 entries, 0 to 2154

Data	columns (total 41 columns	mns):	
#	Column	Non-Null Count	Dtype
0	CustomerID	2155 non-null	object
1	Recency	2155 non-null	int64
2	Frequency	2155 non-null	int64
3	Monetary	2155 non-null	float64
4	R_Score	2155 non-null	int64
5	F_Score	2155 non-null	int64
6	M_Score	2155 non-null	int64
7	RFMScore	2155 non-null	object
8	Customer_Segmentation	2155 non-null	object
9	CompanyName	2155 non-null	object
10	ContactName	2155 non-null	object
11	ContactTitle	2155 non-null	object
12	Address	2155 non-null	object

```
13
                             2155 non-null
                                             object
     City
 14
     Country
                            2155 non-null
                                             object
 15
                            2155 non-null
                                             object
     Phone
     OrderID
                            2155 non-null
                                             int64
 16
 17
     EmployeeID
                            2155 non-null
                                             int64
    OrderDate
                            2155 non-null
                                             object
 18
     RequiredDate
 19
                            2155 non-null
                                             object
     ShipVia
                            2155 non-null
 20
                                             int64
 21
     Freight
                            2155 non-null
                                             object
 22
     ShipName
                            2155 non-null
                                             object
 23
     ShipAddress
                            2155 non-null
                                             object
 24
     ShipCity
                            2155 non-null
 25
     ShipCountry
                            2155 non-null
                                             object
 26
     ProductID
                            2155 non-null
                                             int64
     UnitPrice_x
                            2155 non-null
 27
                                             float32
 28
     Quantity
                            2155 non-null
                                             int64
 29
    Discount
                            2155 non-null
                                             float32
 30
     ProductName
                            2155 non-null
                                             object
 31
     SupplierID
                            2155 non-null
                                             int64
     CategoryID
 32
                            2155 non-null
                                             int64
 33
     QuantityPerUnit
                            2155 non-null
                                             object
 34
     UnitPrice_y
                            2155 non-null
                                             float32
 35
     UnitsInStock
                             2155 non-null
                                             int64
 36
     UnitsOnOrder
                             2155 non-null
                                             int64
 37
     ReorderLevel
                             2155 non-null
                                             int64
 38
    Discontinued
                            2155 non-null
                                             bool
 39
     recentpurchasedate
                            2155 non-null
                                             object
 40
    TotalSales
                            2155 non-null
                                             float64
dtypes: bool(1), float32(3), float64(2), int64(15), object(20)
memory usage: 667.1+ KB
```

#### 1 data.head()

	CustomerID	Recency	Frequency	Monetary	R_Score	F_Score	M_Score	RFMScore
0	ALFKI	27	12	4583.50001	2	2	2	222
11	ALFKI	27	12	4583.50001	2	2	2	222
10	ALFKI	27	12	4583.50001	2	2	2	222
9	ALFKI	27	12	4583.50001	2	2	2	222
7	ALFKI	27	12	4583.50001	2	2	2	222

5 rows × 41 columns

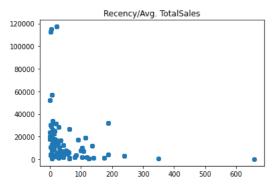


1 data['Customer\_Segmentation'].value\_counts()

Loyal Customers 1066
Gold Customers 459
Silver Customers 318
Low Value Customers 211
Big Spenders 101

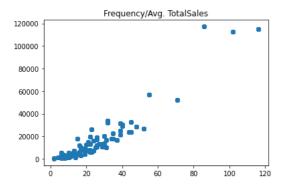
Name: Customer\_Segmentation, dtype: int64

- 1 plt.scatter(data['Recency'], data['Monetary'])
- 2 plt.title('Recency/Avg. TotalSales')
- 3 plt.savefig("RecedncyAvgTotalSales.png")
- 4 plt.show()



1 plt.scatter(data['Frequencv']. data['Monetarv'])

- plt.title('Frequency/Avg. TotalSales')
- plt.savefig("FrequencyAvgTotalSales.png")
  plt.show()



data.to\_csv('./RFM\_Analysis.csv', index=False)

✓ 0s completed at 15:02