

Supermarket_exp

April 19, 2025

```
[9]: import pandas as pd
import numpy as np
import matplotlib as plt
import matplotlib.pyplot as plt
import plotly.express as px # for making it interactive
from sklearn.linear_model import LinearRegression # for prediction of the next
    month revenue
import seaborn as sns
```

```
[7]: data0=pd.read_csv('ORIGINAL SUPERMARKET_DATA.csv')
```

```
[11]: data0.head()
```

```
[11]:
```

	Invoice ID	Branch	City	Customer type	Gender	\
0	750-67-8428	A	Yangon	Member	Female	
1	226-31-3081	C	Naypyitaw	Normal	Female	
2	631-41-3108	A	Yangon	Normal	Male	
3	123-19-1176	A	Yangon	Member	Male	
4	373-73-7910	A	Yangon	Normal	Male	

	Product line	Unit price	Quantity	Tax 5%	Total	Date	\
0	Health and beauty	74.69	7	26.1415	548.9715	1/5/2019	
1	Electronic accessories	15.28	5	3.8200	80.2200	3/8/2019	
2	Home and lifestyle	46.33	7	16.2155	340.5255	3/3/2019	
3	Health and beauty	58.22	8	23.2880	489.0480	1/27/2019	
4	Sports and travel	86.31	7	30.2085	634.3785	2/8/2019	

	Time	Payment	cogs	gross margin percentage	gross income	Rating
0	13:08	Ewallet	522.83	4.761905	26.1415	9.1
1	10:29	Cash	76.40	4.761905	3.8200	9.6
2	13:23	Credit card	324.31	4.761905	16.2155	7.4
3	20:33	Ewallet	465.76	4.761905	23.2880	8.4
4	10:37	Ewallet	604.17	4.761905	30.2085	5.3

```
[12]: data0.columns
```

```
[12]: Index(['Invoice ID', 'Branch', 'City', 'Customer type', 'Gender',
          'Product line', 'Unit price', 'Quantity', 'Tax 5%', 'Total', 'Date',
```

```

        'Time', 'Payment', 'cogs', 'gross margin percentage', 'gross income',
        'Rating'],
        dtype='object')

```

```

[ ]: ### Added new columns to the dataset.csv to keep the original data unchanged

```

```

[12]: data= pd.read_csv('dataset.csv')

data.info()

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 22 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Unnamed: 0.2                          1000 non-null   int64
1   Unnamed: 0.1                          1000 non-null   int64
2   Unnamed: 0                             1000 non-null   int64
3   Invoice ID                             1000 non-null   object
4   Branch                                1000 non-null   object
5   City                                  1000 non-null   object
6   Customer type                         1000 non-null   object
7   Gender                                1000 non-null   object
8   Product line                          1000 non-null   object
9   Unit price                            1000 non-null   float64
10  Quantity                              1000 non-null   int64
11  Tax 5%                                1000 non-null   float64
12  Total                                 1000 non-null   float64
13  Time                                  1000 non-null   object
14  Payment                              1000 non-null   object
15  cogs                                  1000 non-null   float64
16  gross margin percentage               1000 non-null   float64
17  gross income                          1000 non-null   float64
18  Rating                                1000 non-null   float64
19  Day                                   1000 non-null   object
20  Date(dd-mm-yyyy)                     1000 non-null   object
21  Month                                 1000 non-null   object
dtypes: float64(7), int64(4), object(11)
memory usage: 172.0+ KB

```

```

[ ]: data['Date(dd-mm-yyyy)']=pd.to_datetime(data['Date(dd-mm-yyyy)'],
↳format='%d-%m-%Y' )
data['Day']=data['Date(dd-mm-yyyy)'].dt.day_name()

```

```

[14]: data.describe()

```

```
[14]:      Unnamed: 0.2  Unnamed: 0.1  Unnamed: 0  Unit price  Quantity \
count    1000.000000    1000.000000    1000.000000    1000.000000    1000.000000
mean      499.500000      499.500000      499.500000      55.672130      5.510000
std       288.819436      288.819436      288.819436      26.494628      2.923431
min         0.000000         0.000000         0.000000      10.080000      1.000000
25%       249.750000      249.750000      249.750000      32.875000      3.000000
50%       499.500000      499.500000      499.500000      55.230000      5.000000
75%       749.250000      749.250000      749.250000      77.935000      8.000000
max       999.000000      999.000000      999.000000      99.960000     10.000000
```

```
      Tax 5%      Total      cogs  gross margin percentage \
count    1000.000000    1000.000000    1000.000000      1000.000000
mean      15.379369      322.966749      307.58738      4.761905
std       11.708825      245.885335      234.17651      0.000000
min         0.508500      10.678500      10.17000      4.761905
25%         5.924875      124.422375      118.49750      4.761905
50%        12.088000      253.848000      241.76000      4.761905
75%        22.445250      471.350250      448.90500      4.761905
max        49.650000     1042.650000      993.00000      4.761905
```

```
      gross income      Rating
count    1000.000000    1000.000000
mean      15.379369      6.97270
std       11.708825      1.71858
min         0.508500      4.00000
25%         5.924875      5.50000
50%        12.088000      7.00000
75%        22.445250      8.50000
max        49.650000     10.00000
```

```
[15]: data.dtypes
```

```
[15]: Unnamed: 0.2      int64
      Unnamed: 0.1      int64
      Unnamed: 0      int64
      Invoice ID      object
      Branch      object
      City      object
      Customer type      object
      Gender      object
      Product line      object
      Unit price      float64
      Quantity      int64
      Tax 5%      float64
      Total      float64
      Time      object
      Payment      object
```

```

cogs                                float64
gross margin percentage             float64
gross income                        float64
Rating                             float64
Day                                 object
Date(dd-mm-yyyy)                   object
Month                               object
dtype: object

```

```
[16]: data['Quantity'].max()
```

```
[16]: np.int64(10)
```

```
[17]: data['Quantity'].min()
```

```
[17]: np.int64(1)
```

```
[18]: data.groupby('Quantity')['Customer type'].apply(list)
```

```

[18]: Quantity
1      [Member, Normal, Normal, Normal, Normal, Norma...
2      [Member, Normal, Normal, Member, Normal, Norma...
3      [Normal, Member, Normal, Member, Member, Norma...
4      [Member, Member, Normal, Member, Member, Membe...
5      [Normal, Normal, Member, Member, Normal, Norma...
6      [Member, Normal, Normal, Normal, Member, Membe...
7      [Member, Normal, Normal, Member, Member, Membe...
8      [Member, Member, Normal, Normal, Normal, Norma...
9      [Normal, Normal, Normal, Member, Member, Norma...
10     [Normal, Normal, Member, Member, Member, Norma...
Name: Customer type, dtype: object

```

1 HOW MANY MEMBERS in the dataset purchased the product where the Quantity is 6??

```
[19]: data.groupby('Quantity')['Customer type'].count()
```

```

[19]: Quantity
1      112
2       91
3       90
4      109
5      102
6       98
7      102
8       85

```

```
9      92
10     119
Name: Customer type, dtype: int64
```

```
[20]: data.groupby('Quantity')['Customer type'].size()
```

```
[20]: Quantity
1      112
2       91
3       90
4      109
5      102
6       98
7      102
8       85
9       92
10     119
Name: Customer type, dtype: int64
```

```
[21]: # need to declare a variable which will store the grouped values
group= data.groupby(['Quantity' , 'Customer type']).size()

group.loc[ 6 , 'Member']
```

```
[21]: np.int64(42)
```

```
[22]: data.groupby(['Quantity' , 'Customer type']).size()
```

```
[22]: Quantity  Customer type
1      Member      57
      Normal      55
2      Member      43
      Normal      48
3      Member      44
      Normal      46
4      Member      61
      Normal      48
5      Member      48
      Normal      54
6      Member      42
      Normal      56
7      Member      51
      Normal      51
8      Member      41
      Normal      44
9      Member      51
      Normal      41
```

```

10      Member      63
      Normal      56
dtype: int64

```

1.1 now checking types of product and their count

```
[23]: data.groupby(['Product line', 'Quantity']).size()
```

```

[23]: Product line      Quantity
      Electronic accessories 1      20
                                   2      8
                                   3     16
                                   4     19
                                   5     17
                                   6     19
                                   7     16
                                   8     17
                                   9     16
                                   10    22
      Fashion accessories 1      30
                                   2     22
                                   3     14
                                   4     21
                                   5     15
                                   6      8
                                   7     21
                                   8     12
                                   9     14
                                   10    21
      Food and beverages 1      15
                                   2     16
                                   3     23
                                   4     18
                                   5     21
                                   6     17
                                   7     12
                                   8     15
                                   9     17
                                   10    20
      Health and beauty 1      15
                                   2     13
                                   3     13
                                   4     15
                                   5     18
                                   6     14
                                   7     19
                                   8     15

```

	9	13
	10	17
Home and lifestyle	1	13
	2	14
	3	13
	4	22
	5	13
	6	21
	7	12
	8	17
	9	18
	10	17
Sports and travel	1	19
	2	18
	3	11
	4	14
	5	18
	6	19
	7	22
	8	9
	9	14
	10	22

dtype: int64

1.2 no of people who brought electronic items

```
[24]: #Storing this data
product=data.groupby(['Product line','Quantity']).size()

product.loc['Electronic accessories'].sum()
```

```
[24]: np.int64(170)
```

2 SORTING DATA ACCORDING TO THE MONTH

```
[25]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 22 columns):
#   Column              Non-Null Count  Dtype
---  -
0   Unnamed: 0.2        1000 non-null  int64
1   Unnamed: 0.1        1000 non-null  int64
2   Unnamed: 0          1000 non-null  int64
3   Invoice ID          1000 non-null  object
4   Branch              1000 non-null  object
```

```

5   City                                1000 non-null  object
6   Customer type                       1000 non-null  object
7   Gender                             1000 non-null  object
8   Product line                        1000 non-null  object
9   Unit price                          1000 non-null  float64
10  Quantity                           1000 non-null  int64
11  Tax 5%                             1000 non-null  float64
12  Total                              1000 non-null  float64
13  Time                               1000 non-null  object
14  Payment                            1000 non-null  object
15  cogs                               1000 non-null  float64
16  gross margin percentage             1000 non-null  float64
17  gross income                       1000 non-null  float64
18  Rating                             1000 non-null  float64
19  Day                                1000 non-null  object
20  Date(dd-mm-yyyy)                   1000 non-null  object
21  Month                              1000 non-null  object
dtypes: float64(7), int64(4), object(11)
memory usage: 172.0+ KB

```

```
[26]: data['Date(dd-mm-yyyy)'].head()
```

```

[26]: 0      5/1/2019
      1      3/3/2019
      2     27-01-2019
      3      8/2/2019
      4     25-02-2019
Name: Date(dd-mm-yyyy), dtype: object

```

```

[ ]: #making a new column for the date format which stores the correct format
#https://docs.python.org/3/library/datetime.html#strftime-and-strptime-behavior
#for using dt we need to convert the column into date time

data['Date']=pd.to_datetime(data['Date'])

# created a new column as Date(dd-mm-yyyy)
data['Date(dd-mm-yyyy)']=data['Date'].dt.strftime('%d-%m-%Y')

```

2.1 need to convert this date format which is in YYYY-MM-DD to DD-MM-YYY

```
[27]: data.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 22 columns):
#   Column                                Non-Null Count  Dtype

```



```

---      -----      -----      -----
0  Unnamed: 0.2      1000 non-null  int64
1  Unnamed: 0.1      1000 non-null  int64
2  Unnamed: 0        1000 non-null  int64
3  Invoice ID         1000 non-null  object
4  Branch             1000 non-null  object
5  City               1000 non-null  object
6  Customer type      1000 non-null  object
7  Gender             1000 non-null  object
8  Product line       1000 non-null  object
9  Unit price         1000 non-null  float64
10 Quantity          1000 non-null  int64
11 Tax 5%            1000 non-null  float64
12 Total             1000 non-null  float64
13 Time              1000 non-null  object
14 Payment           1000 non-null  object
15 cogs              1000 non-null  float64
16 gross margin percentage 1000 non-null  float64
17 gross income       1000 non-null  float64
18 Rating            1000 non-null  float64
19 Day               1000 non-null  object
20 Date(dd-mm-yyyy)  1000 non-null  object
21 Month             1000 non-null  object

```

dtypes: float64(7), int64(4), object(11)

memory usage: 172.0+ KB

[28]: *#ok so new column is added*

```
data.head(20)
```

```

[28]:      Unnamed: 0.2  Unnamed: 0.1  Unnamed: 0  Invoice ID  Branch  City  \
0                0                0            0  750-67-8428      A  Yangon
1                1                1            1  631-41-3108      A  Yangon
2                2                2            2  123-19-1176      A  Yangon
3                3                3            3  373-73-7910      A  Yangon
4                4                4            4  355-53-5943      A  Yangon
5                5                5            5  665-32-9167      A  Yangon
6                6                6            6  365-64-0515      A  Yangon
7                7                7            7  252-56-2699      A  Yangon
8                8                8            8  829-34-3910      A  Yangon
9                9                9            9  656-95-9349      A  Yangon
10               10               10           10  765-26-6951      A  Yangon
11               11               11           11  329-62-1586      A  Yangon
12               12               12           12  636-48-8204      A  Yangon
13               13               13           13  549-59-1358      A  Yangon
14               14               14           14  227-03-5010      A  Yangon
15               15               15           15  189-17-4241      A  Yangon

```

16	16	16	16	848-62-7243	A	Yangon
17	17	17	17	595-11-5460	A	Yangon
18	18	18	18	129-29-8530	A	Yangon
19	19	19	19	272-65-1806	A	Yangon

	Customer type	Gender	Product line	Unit price	...	Total \
0	Member	Female	Health and beauty	74.69	...	548.9715
1	Normal	Male	Home and lifestyle	46.33	...	340.5255
2	Member	Male	Health and beauty	58.22	...	489.0480
3	Normal	Male	Sports and travel	86.31	...	634.3785
4	Member	Female	Electronic accessories	68.84	...	433.6920
5	Member	Female	Health and beauty	36.26	...	76.1460
6	Normal	Female	Electronic accessories	46.95	...	246.4875
7	Normal	Male	Food and beverages	43.19	...	453.4950
8	Normal	Female	Health and beauty	71.38	...	749.4900
9	Member	Female	Health and beauty	68.93	...	506.6355
10	Normal	Male	Sports and travel	72.61	...	457.4430
11	Normal	Male	Food and beverages	54.67	...	172.2105
12	Normal	Male	Electronic accessories	34.56	...	181.4400
13	Member	Male	Sports and travel	88.63	...	279.1845
14	Member	Female	Home and lifestyle	52.59	...	441.7560
15	Normal	Female	Fashion accessories	87.67	...	184.1070
16	Normal	Male	Health and beauty	24.89	...	235.2105
17	Normal	Male	Health and beauty	96.58	...	202.8180
18	Member	Male	Sports and travel	62.62	...	328.7550
19	Normal	Female	Electronic accessories	60.88	...	575.3160

	Time	Payment	cogs	gross margin percentage	gross income \
0	13:08:00	Ewallet	522.83	4.761905	26.1415
1	13:23:00	Credit card	324.31	4.761905	16.2155
2	20:33:00	Ewallet	465.76	4.761905	23.2880
3	10:37:00	Ewallet	604.17	4.761905	30.2085
4	14:36:00	Ewallet	413.04	4.761905	20.6520
5	17:15:00	Credit card	72.52	4.761905	3.6260
6	10:25:00	Ewallet	234.75	4.761905	11.7375
7	16:48:00	Ewallet	431.90	4.761905	21.5950
8	19:21:00	Cash	713.80	4.761905	35.6900
9	11:03:00	Credit card	482.51	4.761905	24.1255
10	10:39:00	Credit card	435.66	4.761905	21.7830
11	18:00:00	Credit card	164.01	4.761905	8.2005
12	11:15:00	Ewallet	172.80	4.761905	8.6400
13	17:36:00	Ewallet	265.89	4.761905	13.2945
14	19:20:00	Credit card	420.72	4.761905	21.0360
15	12:17:00	Credit card	175.34	4.761905	8.7670
16	15:36:00	Cash	224.01	4.761905	11.2005
17	10:12:00	Credit card	193.16	4.761905	9.6580
18	19:15:00	Ewallet	313.10	4.761905	15.6550

19	17:17:00	Ewallet	547.92	4.761905	27.3960
----	----------	---------	--------	----------	---------

	Rating	Day	Date(dd-mm-yyyy)	Month
0	9.1	Saturday	5/1/2019	January
1	7.4	Sunday	3/3/2019	March
2	8.4	Sunday	27-01-2019	January
3	5.3	Friday	8/2/2019	February
4	5.8	Monday	25-02-2019	February
5	7.2	Thursday	10/1/2019	January
6	7.1	Tuesday	12/2/2019	February
7	8.2	Thursday	7/2/2019	February
8	5.7	Friday	29-03-2019	March
9	4.6	Monday	11/3/2019	March
10	6.9	Tuesday	1/1/2019	January
11	8.6	Monday	21-01-2019	January
12	9.9	Sunday	17-02-2019	February
13	6.0	Saturday	2/3/2019	March
14	8.5	Friday	22-03-2019	March
15	7.7	Sunday	10/3/2019	March
16	7.4	Friday	15-03-2019	March
17	5.1	Friday	15-03-2019	March
18	7.0	Sunday	10/3/2019	March
19	4.7	Tuesday	15-01-2019	January

[20 rows x 22 columns]

[29]: *#ok so now we have the correct formatted data so we don't need the old column so
→ we are gonna delete that*

data.head()

[29]:

	Unnamed: 0.2	Unnamed: 0.1	Unnamed: 0	Invoice ID	Branch	City	\
0	0	0	0	750-67-8428	A	Yangon	
1	1	1	1	631-41-3108	A	Yangon	
2	2	2	2	123-19-1176	A	Yangon	
3	3	3	3	373-73-7910	A	Yangon	
4	4	4	4	355-53-5943	A	Yangon	

	Customer type	Gender	Product line	Unit price	...	Total	\
0	Member	Female	Health and beauty	74.69	...	548.9715	
1	Normal	Male	Home and lifestyle	46.33	...	340.5255	
2	Member	Male	Health and beauty	58.22	...	489.0480	
3	Normal	Male	Sports and travel	86.31	...	634.3785	
4	Member	Female	Electronic accessories	68.84	...	433.6920	

	Time	Payment	cogs	gross margin percentage	gross income	\
0	13:08:00	Ewallet	522.83	4.761905	26.1415	

1	13:23:00	Credit card	324.31	4.761905	16.2155
2	20:33:00	Ewallet	465.76	4.761905	23.2880
3	10:37:00	Ewallet	604.17	4.761905	30.2085
4	14:36:00	Ewallet	413.04	4.761905	20.6520

	Rating	Day	Date(dd-mm-yyyy)	Month
0	9.1	Saturday	5/1/2019	January
1	7.4	Sunday	3/3/2019	March
2	8.4	Sunday	27-01-2019	January
3	5.3	Friday	8/2/2019	February
4	5.8	Monday	25-02-2019	February

[5 rows x 22 columns]

```
[ ]: #now sorting according to month
#need to create a column called month
data['Date(dd-mm-yyyy)']=pd.to_datetime(data['Date(dd-mm-yyyy)'],
    ↪format='%d-%m-%Y')

data['Month']=data['Date(dd-mm-yyyy)'].dt.month_name()
```

```
[ ]: # Pandas displays datetime objects in the format YYYY-MM-DD by default.

#to to display the date in dd-mm-yyyy format, just need to format it like
    ↪this:

# Using strftime() if you want the format to look nice for display or export.

# Don't use it before extracting month/day, because then it becomes a string
    ↪again

data['Date(dd-mm-yyyy)']=data['Date(dd-mm-yyyy)'].dt.strftime('%d-%m-%Y')
```

```
[31]: data.head(5)
```

```
[31]: Unnamed: 0.2 Unnamed: 0.1 Unnamed: 0 Invoice ID Branch City \
0 0 0 0 750-67-8428 A Yangon
1 1 1 1 631-41-3108 A Yangon
2 2 2 2 123-19-1176 A Yangon
3 3 3 3 373-73-7910 A Yangon
4 4 4 4 355-53-5943 A Yangon

Customer type Gender Product line Unit price ... Total \
0 Member Female Health and beauty 74.69 ... 548.9715
1 Normal Male Home and lifestyle 46.33 ... 340.5255
```

2	Member	Male	Health and beauty	58.22	...	489.0480
3	Normal	Male	Sports and travel	86.31	...	634.3785
4	Member	Female	Electronic accessories	68.84	...	433.6920

	Time	Payment	cogs	gross margin percentage	gross income	\
0	13:08:00	Ewallet	522.83	4.761905	26.1415	
1	13:23:00	Credit card	324.31	4.761905	16.2155	
2	20:33:00	Ewallet	465.76	4.761905	23.2880	
3	10:37:00	Ewallet	604.17	4.761905	30.2085	
4	14:36:00	Ewallet	413.04	4.761905	20.6520	

	Rating	Day	Date(dd-mm-yyyy)	Month
0	9.1	Saturday	5/1/2019	January
1	7.4	Sunday	3/3/2019	March
2	8.4	Sunday	27-01-2019	January
3	5.3	Friday	8/2/2019	February
4	5.8	Monday	25-02-2019	February

[5 rows x 22 columns]

```
[32]: #taking the count of sales according to month
      #month order is in alphabetical order
      data.groupby(['Month' , 'Product line']).size()
```

```
[32]: Month      Product line
      February  Electronic accessories    54
           Fashion accessories           60
           Food and beverages            62
           Health and beauty              46
           Home and lifestyle              38
           Sports and travel               43
      January   Electronic accessories    54
           Fashion accessories           64
           Food and beverages            56
           Health and beauty              49
           Home and lifestyle              59
           Sports and travel               70
      March     Electronic accessories    62
           Fashion accessories           54
           Food and beverages            56
           Health and beauty              57
           Home and lifestyle              63
           Sports and travel               53
      dtype: int64
```

```
[33]: data['Month'].count
```

```
[33]: <bound method Series.count of 0      January
      1      March
      2      January
      3      February
      4      February
      ...
      995     January
      996     January
      997      March
      998     February
      999     January
      Name: Month, Length: 1000, dtype: object>
```

```
[34]: data['Month'].nunique
```

```
[34]: <bound method IndexOpsMixin.nunique of 0      January
      1      March
      2      January
      3      February
      4      February
      ...
      995     January
      996     January
      997      March
      998     February
      999     January
      Name: Month, Length: 1000, dtype: object>
```

```
[35]: data['Month'].nunique()
```

```
[35]: 3
```

```
[36]: data.groupby([data['Month']=='January' , data['Product line'] == 'Sports and_
      ↪travel'])
```

```
[36]: <pandas.core.groupby.generic.DataFrameGroupBy object at 0x00000209B0DDD350>
```

3 saving the file

```
[38]: data.to_csv('dataset.csv' , index='false')
```

```
[39]: data.head()
```

```
[39]:   Unnamed: 0.2  Unnamed: 0.1  Unnamed: 0  Invoice ID Branch  City \
0              0              0            0  750-67-8428      A  Yangon
1              1              1            1  631-41-3108      A  Yangon
2              2              2            2  123-19-1176      A  Yangon
```

3	3	3	3	373-73-7910	A	Yangon
4	4	4	4	355-53-5943	A	Yangon

	Customer type	Gender	Product line	Unit price	...	Total	\
0	Member	Female	Health and beauty	74.69	...	548.9715	
1	Normal	Male	Home and lifestyle	46.33	...	340.5255	
2	Member	Male	Health and beauty	58.22	...	489.0480	
3	Normal	Male	Sports and travel	86.31	...	634.3785	
4	Member	Female	Electronic accessories	68.84	...	433.6920	

	Time	Payment	cogs	gross margin	percentage	gross income	\
0	13:08:00	Ewallet	522.83		4.761905	26.1415	
1	13:23:00	Credit card	324.31		4.761905	16.2155	
2	20:33:00	Ewallet	465.76		4.761905	23.2880	
3	10:37:00	Ewallet	604.17		4.761905	30.2085	
4	14:36:00	Ewallet	413.04		4.761905	20.6520	

	Rating	Day	Date(dd-mm-yyyy)	Month
0	9.1	Saturday	5/1/2019	January
1	7.4	Sunday	3/3/2019	March
2	8.4	Sunday	27-01-2019	January
3	5.3	Friday	8/2/2019	February
4	5.8	Monday	25-02-2019	February

[5 rows x 22 columns]

```
[40]: data['Product line'].unique()
```

```
[40]: array(['Health and beauty', 'Home and lifestyle', 'Sports and travel',
        'Electronic accessories', 'Food and beverages',
        'Fashion accessories'], dtype=object)
```

```
[41]: data['Product line'].nunique()
```

```
[41]: 6
```

```
[42]: data.groupby(['Month', 'Product line']).size()
```

```
[42]: Month    Product line
February Electronic accessories    54
          Fashion accessories      60
          Food and beverages      62
          Health and beauty        46
          Home and lifestyle       38
          Sports and travel        43
January   Electronic accessories    54
          Fashion accessories      64
```

	Food and beverages	56
	Health and beauty	49
	Home and lifestyle	59
	Sports and travel	70
March	Electronic accessories	62
	Fashion accessories	54
	Food and beverages	56
	Health and beauty	57
	Home and lifestyle	63
	Sports and travel	53

dtype: int64

```
[43]: data.groupby(['Month' , 'Product line' , 'Quantity']).size()
```

```
[43]: Month      Product line      Quantity
February  Electronic accessories  1          10
                                                2           2
                                                3           3
                                                4           2
                                                5           5
                                                ..
March      Sports and travel      6           6
                                                7           7
                                                8           4
                                                9           5
                                                10          8

Length: 180, dtype: int64
```

```
[44]: data.head(10)
```

```
[44]: Unnamed: 0.2  Unnamed: 0.1  Unnamed: 0  Invoice ID Branch  City \
0              0              0           0  750-67-8428    A  Yangon
1              1              1           1  631-41-3108    A  Yangon
2              2              2           2  123-19-1176    A  Yangon
3              3              3           3  373-73-7910    A  Yangon
4              4              4           4  355-53-5943    A  Yangon
5              5              5           5  665-32-9167    A  Yangon
6              6              6           6  365-64-0515    A  Yangon
7              7              7           7  252-56-2699    A  Yangon
8              8              8           8  829-34-3910    A  Yangon
9              9              9           9  656-95-9349    A  Yangon

Customer type  Gender      Product line  Unit price  ...      Total  \
0      Member  Female      Health and beauty      74.69  ...  548.9715
1      Normal   Male      Home and lifestyle      46.33  ...  340.5255
2      Member   Male      Health and beauty      58.22  ...  489.0480
3      Normal   Male      Sports and travel      86.31  ...  634.3785
```


4	Member	Female	Electronic accessories	68.84	...	433.6920
5	Member	Female	Health and beauty	36.26	...	76.1460
6	Normal	Female	Electronic accessories	46.95	...	246.4875
7	Normal	Male	Food and beverages	43.19	...	453.4950
8	Normal	Female	Health and beauty	71.38	...	749.4900
9	Member	Female	Health and beauty	68.93	...	506.6355

	Time	Payment	cogs	gross margin percentage	gross income \
0	13:08:00	Ewallet	522.83	4.761905	26.1415
1	13:23:00	Credit card	324.31	4.761905	16.2155
2	20:33:00	Ewallet	465.76	4.761905	23.2880
3	10:37:00	Ewallet	604.17	4.761905	30.2085
4	14:36:00	Ewallet	413.04	4.761905	20.6520
5	17:15:00	Credit card	72.52	4.761905	3.6260
6	10:25:00	Ewallet	234.75	4.761905	11.7375
7	16:48:00	Ewallet	431.90	4.761905	21.5950
8	19:21:00	Cash	713.80	4.761905	35.6900
9	11:03:00	Credit card	482.51	4.761905	24.1255

	Rating	Day	Date(dd-mm-yyyy)	Month
0	9.1	Saturday	5/1/2019	January
1	7.4	Sunday	3/3/2019	March
2	8.4	Sunday	27-01-2019	January
3	5.3	Friday	8/2/2019	February
4	5.8	Monday	25-02-2019	February
5	7.2	Thursday	10/1/2019	January
6	7.1	Tuesday	12/2/2019	February
7	8.2	Thursday	7/2/2019	February
8	5.7	Friday	29-03-2019	March
9	4.6	Monday	11/3/2019	March

[10 rows x 22 columns]

```
[45]: x=data.groupby(['Month' , 'Product line']).size()
```

```
x
```

```
[45]: Month      Product line
February  Electronic accessories    54
          Fashion accessories       60
          Food and beverages        62
          Health and beauty         46
          Home and lifestyle        38
          Sports and travel         43
January   Electronic accessories    54
          Fashion accessories       64
          Food and beverages        56
```

	Health and beauty	49
	Home and lifestyle	59
	Sports and travel	70
March	Electronic accessories	62
	Fashion accessories	54
	Food and beverages	56
	Health and beauty	57
	Home and lifestyle	63
	Sports and travel	53

dtype: int64

```
[46]: #size() = Compute the number of values in each group

y=data.groupby(['Product line' , 'Quantity']).size()
y
```

```
[46]: Product line      Quantity
Electronic accessories 1         20
                        2          8
                        3         16
                        4         19
                        5         17
                        6         19
                        7         16
                        8         17
                        9         16
                        10        22
Fashion accessories    1         30
                        2         22
                        3         14
                        4         21
                        5         15
                        6          8
                        7         21
                        8         12
                        9         14
                        10        21
Food and beverages     1         15
                        2         16
                        3         23
                        4         18
                        5         21
                        6         17
                        7         12
                        8         15
                        9         17
                        10        20
```

Health and beauty	1	15
	2	13
	3	13
	4	15
	5	18
	6	14
	7	19
	8	15
	9	13
	10	17
Home and lifestyle	1	13
	2	14
	3	13
	4	22
	5	13
	6	21
	7	12
	8	17
	9	18
	10	17
Sports and travel	1	19
	2	18
	3	11
	4	14
	5	18
	6	19
	7	22
	8	9
	9	14
	10	22

dtype: int64

```
[47]: data['Quantity'].sum()
```

```
[47]: np.int64(5510)
```

```
[48]: pd.concat([x ,y],axis=0).head(10)
```

```
[48]: February  Electronic accessories  54
      Fashion accessories  60
      Food and beverages  62
      Health and beauty  46
      Home and lifestyle  38
      Sports and travel  43
      January  Electronic accessories  54
      Fashion accessories  64
      Food and beverages  56
```

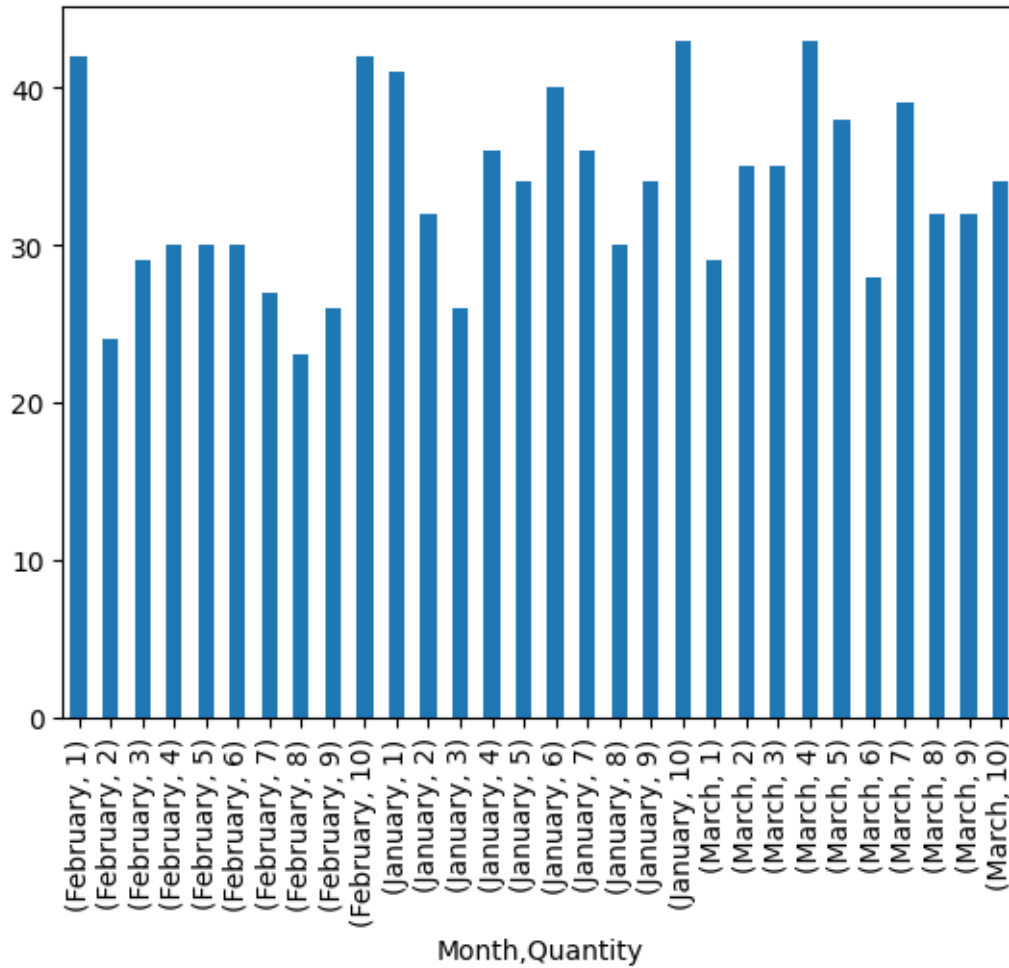
dtype: int64

```
[49]: z=data.groupby(['Month' , 'Quantity']).size()
      z
```

```
[49]: Month      Quantity
      February  1          42
              2          24
              3          29
              4          30
              5          30
              6          30
              7          27
              8          23
              9          26
             10          42
      January   1          41
              2          32
              3          26
              4          36
              5          34
              6          40
              7          36
              8          30
              9          34
             10          43
      March     1          29
              2          35
              3          35
              4          43
              5          38
              6          28
              7          39
              8          32
              9          32
             10          34
      dtype: int64
```

```
[50]: z.plot.bar()
```

```
[50]: <Axes: xlabel='Month,Quantity'>
```



```
[ ]: ### Need to make this graph more visually appealing
```

```
[51]: data.groupby(['Month' , 'Date(dd-mm-yyyy)', 'Product line' , 'Quantity' ]).
      ↪size().reset_index(name='Count').head(50)
```

```
[51]:
```

	Month	Date(dd-mm-yyyy)	Product line	Quantity	Count
0	February	1/2/2019	Electronic accessories	9	1
1	February	1/2/2019	Food and beverages	5	1
2	February	1/2/2019	Food and beverages	6	1
3	February	1/2/2019	Health and beauty	4	1
4	February	1/2/2019	Home and lifestyle	9	1
5	February	1/2/2019	Sports and travel	7	1
6	February	10/2/2019	Electronic accessories	3	1
7	February	10/2/2019	Electronic accessories	6	1
8	February	10/2/2019	Electronic accessories	10	1
9	February	10/2/2019	Fashion accessories	7	2

10	February	10/2/2019	Fashion accessories	10	1
11	February	10/2/2019	Food and beverages	5	1
12	February	10/2/2019	Health and beauty	2	1
13	February	10/2/2019	Home and lifestyle	1	1
14	February	10/2/2019	Home and lifestyle	7	1
15	February	10/2/2019	Sports and travel	4	1
16	February	11/2/2019	Fashion accessories	10	1
17	February	11/2/2019	Food and beverages	10	2
18	February	11/2/2019	Health and beauty	9	1
19	February	11/2/2019	Home and lifestyle	3	2
20	February	11/2/2019	Home and lifestyle	5	1
21	February	11/2/2019	Sports and travel	4	1
22	February	12/2/2019	Electronic accessories	1	1
23	February	12/2/2019	Electronic accessories	5	1
24	February	12/2/2019	Fashion accessories	1	1
25	February	12/2/2019	Fashion accessories	8	1
26	February	12/2/2019	Food and beverages	10	1
27	February	12/2/2019	Health and beauty	6	1
28	February	12/2/2019	Home and lifestyle	10	1
29	February	12/2/2019	Sports and travel	2	1
30	February	13-02-2019	Electronic accessories	1	1
31	February	13-02-2019	Electronic accessories	4	1
32	February	13-02-2019	Fashion accessories	2	1
33	February	13-02-2019	Fashion accessories	5	1
34	February	13-02-2019	Fashion accessories	7	1
35	February	13-02-2019	Food and beverages	2	1
36	February	13-02-2019	Sports and travel	4	1
37	February	13-02-2019	Sports and travel	6	1
38	February	14-02-2019	Electronic accessories	8	1
39	February	14-02-2019	Food and beverages	2	1
40	February	14-02-2019	Food and beverages	10	1
41	February	14-02-2019	Health and beauty	1	1
42	February	14-02-2019	Health and beauty	2	1
43	February	14-02-2019	Health and beauty	4	1
44	February	14-02-2019	Sports and travel	3	1
45	February	14-02-2019	Sports and travel	7	1
46	February	15-02-2019	Electronic accessories	1	1
47	February	15-02-2019	Electronic accessories	9	1
48	February	15-02-2019	Electronic accessories	10	1
49	February	15-02-2019	Fashion accessories	3	2

```
[53]: GROUP_1_DATA=data.groupby(['Month' , 'Date(dd-mm-yyyy)', 'Day', 'Product line' ,
↳ 'Quantity' ]).size().reset_index(name='Count')

GROUP_1_DATA.to_csv('GROUP_1_DATA .csv' , index= False )
```

```
[54]: data2=pd.read_csv('GROUP_1_DATA .csv')

data2.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 903 entries, 0 to 902
Data columns (total 6 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Month                 903 non-null   object
1   Date(dd-mm-yyyy)     903 non-null   object
2   Day                  903 non-null   object
3   Product line         903 non-null   object
4   Quantity             903 non-null   int64
5   Count               903 non-null   int64
dtypes: int64(2), object(4)
memory usage: 42.5+ KB
```

```
[55]: data2['Month'].unique()
```

```
[55]: array(['February', 'January', 'March'], dtype=object)
```

```
[56]: data2['Product line'].unique()
```

```
[56]: array(['Electronic accessories', 'Food and beverages',
           'Health and beauty', 'Home and lifestyle', 'Sports and travel',
           'Fashion accessories'], dtype=object)
```

```
[57]: data2.groupby(['Month', 'Day', 'Product line'])['Quantity'].sum().
      ↪reset_index(name='Count').head(20)
```

```
[57]:
```

	Month	Day	Product line	Count
0	February	Friday	Electronic accessories	39
1	February	Friday	Fashion accessories	56
2	February	Friday	Food and beverages	24
3	February	Friday	Health and beauty	36
4	February	Friday	Home and lifestyle	38
5	February	Friday	Sports and travel	55
6	February	Monday	Electronic accessories	31
7	February	Monday	Fashion accessories	35
8	February	Monday	Food and beverages	23
9	February	Monday	Health and beauty	39
10	February	Monday	Home and lifestyle	29
11	February	Monday	Sports and travel	15
12	February	Saturday	Electronic accessories	44
13	February	Saturday	Fashion accessories	54
14	February	Saturday	Food and beverages	56
15	February	Saturday	Health and beauty	21

16	February	Saturday	Home and lifestyle	20
17	February	Saturday	Sports and travel	35
18	February	Sunday	Electronic accessories	48
19	February	Sunday	Fashion accessories	28

```
[58]: #storing the grouped dataframe in the variable called graph
```

```
graph=data2.groupby(['Month' , 'Day', 'Product line'])['Quantity'].sum().
    ↪reset_index(name='Count')
```

```
graph
```

```
[58]:
```

	Month	Day	Product line	Count
0	February	Friday	Electronic accessories	39
1	February	Friday	Fashion accessories	56
2	February	Friday	Food and beverages	24
3	February	Friday	Health and beauty	36
4	February	Friday	Home and lifestyle	38
..
121	March	Wednesday	Fashion accessories	39
122	March	Wednesday	Food and beverages	37
123	March	Wednesday	Health and beauty	19
124	March	Wednesday	Home and lifestyle	39
125	March	Wednesday	Sports and travel	70

```
[126 rows x 4 columns]
```

```
[59]: graph.columns
```

```
[59]: Index(['Month', 'Day', 'Product line', 'Count'], dtype='object')
```

```
[60]: graph['Month'].unique()
```

```
[60]: array(['February', 'January', 'March'], dtype=object)
```

3.1 PLOTTING THE GRAPH

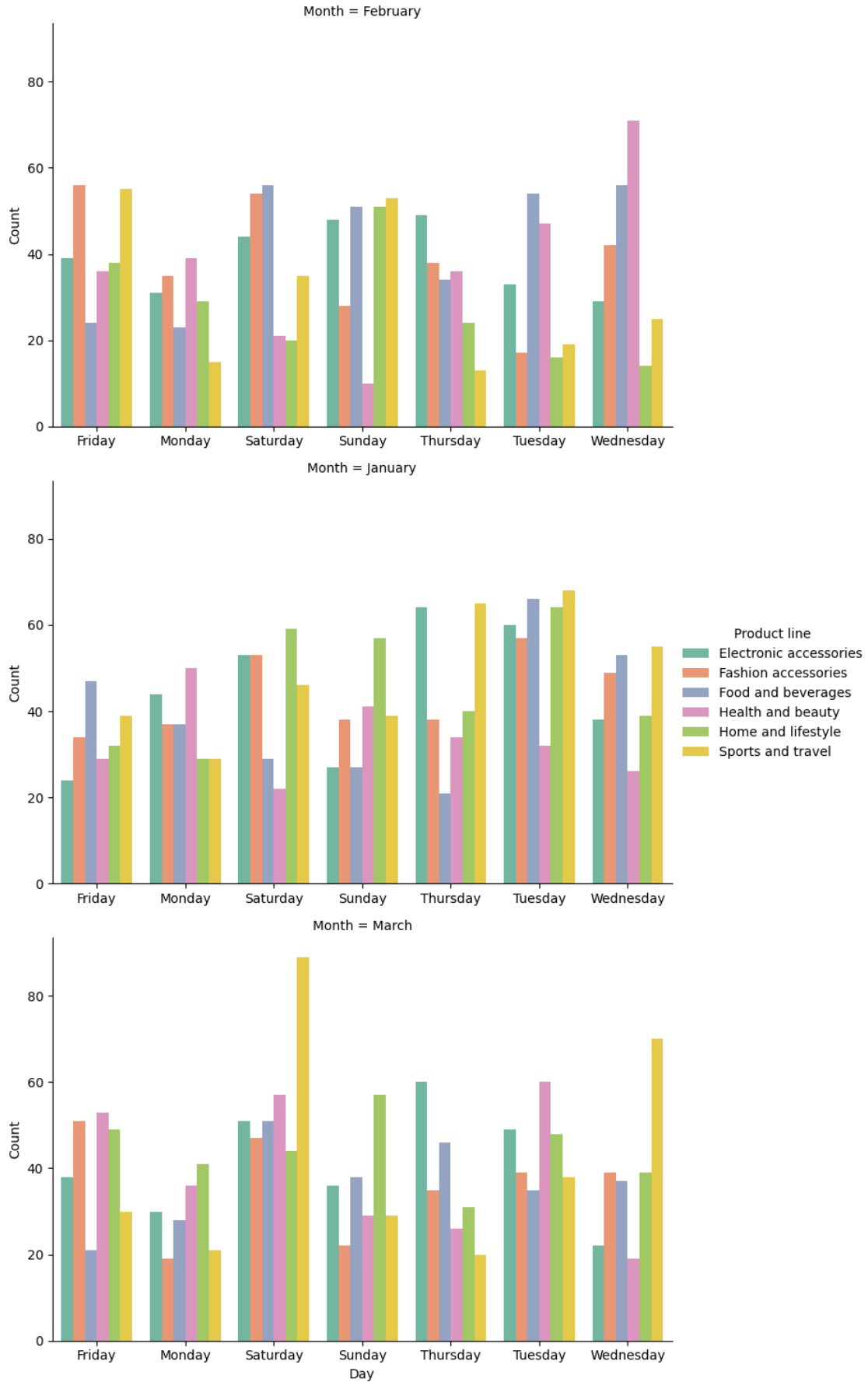
```
[61]: plt.figure(figsize=(20, 10)) #size of figure 20 bredth 10 height
```

```
[61]: <Figure size 2000x1000 with 0 Axes>
```

```
<Figure size 2000x1000 with 0 Axes>
```


4 Quantity of product sold on week days of thr month

```
[63]: g= sns.catplot(data=graph
                    ,x='Day' ,
                    y='Count' ,
                    hue='Product line' ,
                    row='Month' ,
                    kind='bar' ,
                    height=5,
                    aspect=1.5,
                    palette='Set2',
                    sharex=False)
#sharex=false for displaying the days in all the graphs
# change col to row to make it vertical
plt.show()
```



4.0.1 MAKING IT INTERACTIVE USING PLOTLY EXPRESS

```
[64]: g= px.bar(  
    graph,  
    x="Day",  
    y="Count",  
    color="Product line",  
    facet_row="Month",  
    barmode="group",  
    height=1500,  
    color_discrete_sequence=px.colors.qualitative.Set2,  
    title="Sales by Product Line and Day (Interactive)"  
  
    )  
g.update_xaxes(matches='x')  
g.update_yaxes(matches='y')  
g.show()
```

```
[65]: g.for_each_axis(lambda axis :  
    axis.update(title='Day'))  
  
g.update_xaxes(showticklabels=True)
```

```
[66]: data2.columns
```

```
[66]: Index(['Month', 'Date(dd-mm-yyyy)', 'Day', 'Product line', 'Quantity',  
    'Count'],  
    dtype='object')
```

```
[67]: data2['Product line'].unique()
```

```
[67]: array(['Electronic accessories', 'Food and beverages',  
    'Health and beauty', 'Home and lifestyle', 'Sports and travel',  
    'Fashion accessories'], dtype=object)
```

```
[68]: data2['Product line'].nunique()
```

```
[68]: 6
```

```
[69]: data2['Product line'].count()
```

```
[69]: np.int64(903)
```

```
[70]: data2.describe()
```

```
[70]:
```

	Quantity	Count
count	903.000000	903.000000
mean	5.513843	1.107420
std	2.930378	0.333928
min	1.000000	1.000000
25%	3.000000	1.000000
50%	5.000000	1.000000
75%	8.000000	1.000000
max	10.000000	3.000000

```
[71]: data2.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 903 entries, 0 to 902
Data columns (total 6 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Month                 903 non-null   object
1   Date(dd-mm-yyyy)     903 non-null   object
2   Day                   903 non-null   object
3   Product line         903 non-null   object
4   Quantity              903 non-null   int64
5   Count                 903 non-null   int64
dtypes: int64(2), object(4)
memory usage: 42.5+ KB
```

```
[ ]: #data is clean and no missing columns are there in it 903== 903`
```

5 adding tax column to data 2

```
[72]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 22 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Unnamed: 0.2          1000 non-null   int64
1   Unnamed: 0.1          1000 non-null   int64
2   Unnamed: 0            1000 non-null   int64
3   Invoice ID            1000 non-null   object
4   Branch               1000 non-null   object
5   City                 1000 non-null   object
6   Customer type        1000 non-null   object
7   Gender               1000 non-null   object
8   Product line         1000 non-null   object
9   Unit price           1000 non-null   float64
10  Quantity             1000 non-null   int64
```

```

11 Tax 5%                1000 non-null    float64
12 Total                1000 non-null    float64
13 Time                 1000 non-null    object
14 Payment              1000 non-null    object
15 cogs                 1000 non-null    float64
16 gross margin percentage 1000 non-null    float64
17 gross income         1000 non-null    float64
18 Rating               1000 non-null    float64
19 Day                  1000 non-null    object
20 Date(dd-mm-yyyy)     1000 non-null    object
21 Month                1000 non-null    object
dtypes: float64(7), int64(4), object(11)
memory usage: 172.0+ KB

```

```
[73]: data.columns
```

```
[73]: Index(['Unnamed: 0.2', 'Unnamed: 0.1', 'Unnamed: 0', 'Invoice ID', 'Branch',
          'City', 'Customer type', 'Gender', 'Product line', 'Unit price',
          'Quantity', 'Tax 5%', 'Total', 'Time', 'Payment', 'cogs',
          'gross margin percentage', 'gross income', 'Rating', 'Day',
          'Date(dd-mm-yyyy)', 'Month'],
          dtype='object')
```

```
[75]: data2['Tax']=data['Tax 5%'].reindex(data2.index) # in data 2 tax column will
        contain the values from the column tax 5% in data
```

```
[76]: data2.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 903 entries, 0 to 902
Data columns (total 7 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Month                  903 non-null    object
1   Date(dd-mm-yyyy)      903 non-null    object
2   Day                    903 non-null    object
3   Product line          903 non-null    object
4   Quantity               903 non-null    int64
5   Count                 903 non-null    int64
6   Tax                    903 non-null    float64
dtypes: float64(1), int64(2), object(4)
memory usage: 49.5+ KB

```

```
[77]: data2.describe()
```

```

[77]:      Quantity      Count      Tax
count  903.000000  903.000000  903.000000
mean    5.513843    1.107420   15.303181

```

std	2.930378	0.333928	11.657250
min	1.000000	1.000000	0.604500
25%	3.000000	1.000000	5.966000
50%	5.000000	1.000000	12.048000
75%	8.000000	1.000000	22.258500
max	10.000000	3.000000	49.650000

```
[78]: data2.head(10)
```

```
[78]:      Month Date(dd-mm-yyyy)   Day   Product line  Quantity  Count \
0  February      1/2/2019  Friday  Electronic accessories      9      1
1  February      1/2/2019  Friday    Food and beverages      5      1
2  February      1/2/2019  Friday    Food and beverages      6      1
3  February      1/2/2019  Friday    Health and beauty      4      1
4  February      1/2/2019  Friday    Home and lifestyle      9      1
5  February      1/2/2019  Friday    Sports and travel      7      1
6  February     10/2/2019  Sunday  Electronic accessories      3      1
7  February     10/2/2019  Sunday  Electronic accessories      6      1
8  February     10/2/2019  Sunday  Electronic accessories     10      1
9  February     10/2/2019  Sunday    Fashion accessories      7      2
```

	Tax
0	26.1415
1	16.2155
2	23.2880
3	30.2085
4	20.6520
5	3.6260
6	11.7375
7	21.5950
8	35.6900
9	24.1255

```
[79]: data2.columns
```

```
[79]: Index(['Month', 'Date(dd-mm-yyyy)', 'Day', 'Product line', 'Quantity', 'Count',
        'Tax'],
        dtype='object')
```

```
[80]: data2.columns.tolist()
```

```
[80]: ['Month',
        'Date(dd-mm-yyyy)',
        'Day',
        'Product line',
        'Quantity',
        'Count',
```

```
'Tax']
```

```
[81]: data2.columns
```

```
[81]: Index(['Month', 'Date(dd-mm-yyyy)', 'Day', 'Product line', 'Quantity', 'Count',  
        'Tax'],  
        dtype='object')
```

```
[82]: data2.drop('Tax' , axis=1, ) # axis 1 cuz its column and if it would have been  
    ↪ rows the axis =0
```

```
[82]:
```

	Month	Date(dd-mm-yyyy)	Day	Product line	Quantity \
0	February	1/2/2019	Friday	Electronic accessories	9
1	February	1/2/2019	Friday	Food and beverages	5
2	February	1/2/2019	Friday	Food and beverages	6
3	February	1/2/2019	Friday	Health and beauty	4
4	February	1/2/2019	Friday	Home and lifestyle	9
..
898	March	9/3/2019	Saturday	Home and lifestyle	7
899	March	9/3/2019	Saturday	Sports and travel	2
900	March	9/3/2019	Saturday	Sports and travel	5
901	March	9/3/2019	Saturday	Sports and travel	9
902	March	9/3/2019	Saturday	Sports and travel	10

	Count
0	1
1	1
2	1
3	1
4	1
..	...
898	1
899	1
900	1
901	1
902	1

```
[903 rows x 6 columns]
```

```
[85]: data2.to_csv('Tax_dataset.csv' ,index=False)
```

```
[ ]: # need to remove the tax column from this data set cuz the values are not  
    ↪ correct
```

5.0.1 Need to group data so the values matches correctly

[86]: `data2.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 903 entries, 0 to 902
Data columns (total 7 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Month                 903 non-null   object
1   Date(dd-mm-yyyy)     903 non-null   object
2   Day                   903 non-null   object
3   Product line         903 non-null   object
4   Quantity              903 non-null   int64
5   Count                 903 non-null   int64
6   Tax                   903 non-null   float64
dtypes: float64(1), int64(2), object(4)
memory usage: 49.5+ KB
```

[87]: `data.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 22 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Unnamed: 0.2          1000 non-null   int64
1   Unnamed: 0.1          1000 non-null   int64
2   Unnamed: 0            1000 non-null   int64
3   Invoice ID             1000 non-null   object
4   Branch                1000 non-null   object
5   City                  1000 non-null   object
6   Customer type         1000 non-null   object
7   Gender                1000 non-null   object
8   Product line          1000 non-null   object
9   Unit price            1000 non-null   float64
10  Quantity              1000 non-null   int64
11  Tax 5%                1000 non-null   float64
12  Total                 1000 non-null   float64
13  Time                  1000 non-null   object
14  Payment               1000 non-null   object
15  cogs                  1000 non-null   float64
16  gross margin percentage 1000 non-null   float64
17  gross income           1000 non-null   float64
18  Rating                1000 non-null   float64
19  Day                   1000 non-null   object
20  Date(dd-mm-yyyy)     1000 non-null   object
21  Month                 1000 non-null   object
dtypes: float64(7), int64(4), object(11)
```


memory usage: 172.0+ KB

5.0.2 Grouping the columns city , Productline , unit price , Quantity , tax 5% on total , Total ,Month ,

```
[94]: TAX_GROUP=data.groupby(['Month','Product line', 'Quantity' , 'Unit price', 'Tax_5%', 'Total']).size().reset_index(name='Count')
```

```
[95]: TAX_GROUP
```

```
[95]:
```

	Month	Product line	Quantity	Unit price	Tax 5%	\
0	February	Electronic accessories	1	28.96	1.4480	
1	February	Electronic accessories	1	39.48	1.9740	
2	February	Electronic accessories	1	39.75	1.9875	
3	February	Electronic accessories	1	60.30	3.0150	
4	February	Electronic accessories	1	60.95	3.0475	
..	
994	March	Sports and travel	10	44.02	22.0100	
995	March	Sports and travel	10	52.26	26.1300	
996	March	Sports and travel	10	54.55	27.2750	
997	March	Sports and travel	10	69.74	34.8700	
998	March	Sports and travel	10	76.92	38.4600	

	Total	Count
0	30.4080	1
1	41.4540	1
2	41.7375	1
3	63.3150	1
4	63.9975	1
..
994	462.2100	1
995	548.7300	1
996	572.7750	1
997	732.2700	1
998	807.6600	1

[999 rows x 7 columns]

```
[90]: TAX_GROUP.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 999 entries, 0 to 998
Data columns (total 7 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Month            999 non-null    object
1   Product line     999 non-null    object
2   Quantity         999 non-null    int64
```

```

3  Unit price      999 non-null    float64
4  Tax 5%          999 non-null    float64
5  Total          999 non-null    float64
6  Count          999 non-null    int64
dtypes: float64(3), int64(2), object(2)
memory usage: 54.8+ KB

```

```
[96]: TAX_GROUP.to_csv('Tax_dataset.csv', index=False)
```

```
[97]: data3=pd.read_csv('Tax_dataset.csv')
```

```
[93]: data3.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 999 entries, 0 to 998
Data columns (total 7 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Month           999 non-null   object
1   Product line    999 non-null   object
2   Quantity        999 non-null   int64
3   Unit price      999 non-null   float64
4   Tax 5%          999 non-null   float64
5   Total           999 non-null   float64
6   Count           999 non-null   int64
dtypes: float64(3), int64(2), object(2)
memory usage: 54.8+ KB

```

```
[98]: data3.head(50)
```

```

[98]:
      Month  Product line  Quantity  Unit price  Tax 5%  Total \
0  February  Electronic accessories      1     28.96   1.4480  30.4080
1  February  Electronic accessories      1     39.48   1.9740  41.4540
2  February  Electronic accessories      1     39.75   1.9875  41.7375
3  February  Electronic accessories      1     60.30   3.0150  63.3150
4  February  Electronic accessories      1     60.95   3.0475  63.9975
5  February  Electronic accessories      1     62.48   3.1240  65.6040
6  February  Electronic accessories      1     71.95   3.5975  75.5475
7  February  Electronic accessories      1     76.82   3.8410  80.6610
8  February  Electronic accessories      1     98.84   4.9420 103.7820
9  February  Electronic accessories      1     99.69   4.9845 104.6745
10 February  Electronic accessories      2     20.89   2.0890  43.8690
11 February  Electronic accessories      2     46.61   4.6610  97.8810
12 February  Electronic accessories      3     48.09   7.2135 151.4835
13 February  Electronic accessories      3     81.40  12.2100 256.4100
14 February  Electronic accessories      3     94.64  14.1960 298.1160
15 February  Electronic accessories      4     32.25   6.4500 135.4500
16 February  Electronic accessories      4     65.94  13.1880 276.9480

```

17	February	Electronic accessories	5	11.81	2.9525	62.0025
18	February	Electronic accessories	5	12.05	3.0125	63.2625
19	February	Electronic accessories	5	34.56	8.6400	181.4400
20	February	Electronic accessories	5	46.95	11.7375	246.4875
21	February	Electronic accessories	5	86.04	21.5100	451.7100
22	February	Electronic accessories	6	12.45	3.7350	78.4350
23	February	Electronic accessories	6	18.93	5.6790	119.2590
24	February	Electronic accessories	6	35.49	10.6470	223.5870
25	February	Electronic accessories	6	46.02	13.8060	289.9260
26	February	Electronic accessories	6	50.45	15.1350	317.8350
27	February	Electronic accessories	6	68.84	20.6520	433.6920
28	February	Electronic accessories	6	87.45	26.2350	550.9350
29	February	Electronic accessories	6	90.70	27.2100	571.4100
30	February	Electronic accessories	7	25.22	8.8270	185.3670
31	February	Electronic accessories	7	26.26	9.1910	193.0110
32	February	Electronic accessories	7	74.58	26.1030	548.1630
33	February	Electronic accessories	7	92.60	32.4100	680.6100
34	February	Electronic accessories	8	14.96	5.9840	125.6640
35	February	Electronic accessories	8	35.74	14.2960	300.2160
36	February	Electronic accessories	8	40.86	16.3440	343.2240
37	February	Electronic accessories	8	57.91	23.1640	486.4440
38	February	Electronic accessories	8	71.89	28.7560	603.8760
39	February	Electronic accessories	8	85.98	34.3920	722.2320
40	February	Electronic accessories	8	99.56	39.8240	836.3040
41	February	Electronic accessories	9	23.07	10.3815	218.0115
42	February	Electronic accessories	9	69.58	31.3110	657.5310
43	February	Electronic accessories	9	75.59	34.0155	714.3255
44	February	Electronic accessories	9	77.63	34.9335	733.6035
45	February	Electronic accessories	9	88.25	39.7125	833.9625
46	February	Electronic accessories	10	17.42	8.7100	182.9100
47	February	Electronic accessories	10	22.95	11.4750	240.9750
48	February	Electronic accessories	10	24.74	12.3700	259.7700
49	February	Electronic accessories	10	32.80	16.4000	344.4000

Count	
0	1
1	1
2	1
3	1
4	1
5	1
6	1
7	1
8	1
9	1
10	1
11	1

```

12      1
13      1
14      1
15      1
16      1
17      1
18      1
19      1
20      1
21      1
22      1
23      1
24      1
25      1
26      1
27      1
28      1
29      1
30      1
31      1
32      1
33      1
34      1
35      1
36      1
37      1
38      1
39      1
40      1
41      1
42      1
43      1
44      1
45      1
46      1
47      1
48      1
49      1

```

```

[99]: T1=data3.groupby(['Month' , 'Product line'])['Tax 5%'].sum().reset_index()
      tax_graph= px.bar(T1 ,y='Tax 5%' , x='Product line' , color='Product line' ,
      ↪,facet_row='Month' , title='Tax graph' ,height=1000)

      tax_graph.update_xaxes(showticklabels=True)
      tax_graph.show()

```

```

[100]: data.info()

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 22 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Unnamed: 0.2                          1000 non-null   int64
1   Unnamed: 0.1                          1000 non-null   int64
2   Unnamed: 0                             1000 non-null   int64
3   Invoice ID                             1000 non-null   object
4   Branch                                1000 non-null   object
5   City                                  1000 non-null   object
6   Customer type                         1000 non-null   object
7   Gender                                1000 non-null   object
8   Product line                          1000 non-null   object
9   Unit price                            1000 non-null   float64
10  Quantity                              1000 non-null   int64
11  Tax 5%                                1000 non-null   float64
12  Total                                 1000 non-null   float64
13  Time                                  1000 non-null   object
14  Payment                              1000 non-null   object
15  cogs                                 1000 non-null   float64
16  gross margin percentage               1000 non-null   float64
17  gross income                         1000 non-null   float64
18  Rating                               1000 non-null   float64
19  Day                                  1000 non-null   object
20  Date(dd-mm-yyyy)                     1000 non-null   object
21  Month                                1000 non-null   object
dtypes: float64(7), int64(4), object(11)
memory usage: 172.0+ KB

```

5.0.3 rating vs product line which product line has the maximum rating

```
[101]: rating_data=data.groupby(['City','Product line' ,'Rating']).size().reset_index()
rating_data
```

```

[101]:
   City  Product line  Rating  0
0  Mandalay  Electronic accessories  4.0  1
1  Mandalay  Electronic accessories  4.2  1
2  Mandalay  Electronic accessories  4.3  1
3  Mandalay  Electronic accessories  4.8  1
4  Mandalay  Electronic accessories  4.9  1
..  ...
669  Yangon  Sports and travel  9.3  2
670  Yangon  Sports and travel  9.6  3
671  Yangon  Sports and travel  9.7  3
672  Yangon  Sports and travel  9.8  1
673  Yangon  Sports and travel  9.9  1

```

[674 rows x 4 columns]

```
[102]: rating_graph = rating_data.pivot_table( columns='Product line' , index='City' ,  
      ↪ values='Rating' , aggfunc='mean' ).reset_index()
```

```
[103]: rating_graph.columns.name=None
```

```
[104]: rating_graph
```

```
[104]:
```

	City	Electronic accessories	Fashion accessories	Food and beverages	\
0	Mandalay	7.069231	6.629268	6.897222	
1	Naypyitaw	6.887500	7.217073	7.140909	
2	Yangon	6.967647	6.902941	7.195238	
		Health and beauty	Home and lifestyle	Sports and travel	
0		7.035294	6.802857	6.682051	
1		7.102778	7.077419	7.038235	
2		6.931429	7.157143	7.218919	

```
[186]: #Giving errors need to melt this data so it is correct form to plot
```

5.1 DATA MELTING need to covert this table to correct format for plotting the graph

```
[106]: melt = pd.melt(rating_graph , id_vars='City' , var_name='Product line' ,  
      ↪ value_name='Average rating')
```

melt

```
[106]:
```

	City	Product line	Average rating
0	Mandalay	Electronic accessories	7.069231
1	Naypyitaw	Electronic accessories	6.887500
2	Yangon	Electronic accessories	6.967647
3	Mandalay	Fashion accessories	6.629268
4	Naypyitaw	Fashion accessories	7.217073
5	Yangon	Fashion accessories	6.902941
6	Mandalay	Food and beverages	6.897222
7	Naypyitaw	Food and beverages	7.140909
8	Yangon	Food and beverages	7.195238
9	Mandalay	Health and beauty	7.035294
10	Naypyitaw	Health and beauty	7.102778
11	Yangon	Health and beauty	6.931429
12	Mandalay	Home and lifestyle	6.802857
13	Naypyitaw	Home and lifestyle	7.077419
14	Yangon	Home and lifestyle	7.157143
15	Mandalay	Sports and travel	6.682051

```

16 Naypyitaw      Sports and travel      7.038235
17 Yangon        Sports and travel      7.218919

```

```

[107]: RG=px.bar(melt,x='City' , y='Average rating' , color='Product line'
↳,barmode='group', title='Average ratings of Product line' )
RG.show()

```

```

[108]: data.info()

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 22 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Unnamed: 0.2                          1000 non-null   int64
1   Unnamed: 0.1                          1000 non-null   int64
2   Unnamed: 0                             1000 non-null   int64
3   Invoice ID                             1000 non-null   object
4   Branch                                1000 non-null   object
5   City                                   1000 non-null   object
6   Customer type                          1000 non-null   object
7   Gender                                 1000 non-null   object
8   Product line                           1000 non-null   object
9   Unit price                             1000 non-null   float64
10  Quantity                               1000 non-null   int64
11  Tax 5%                                 1000 non-null   float64
12  Total                                  1000 non-null   float64
13  Time                                   1000 non-null   object
14  Payment                                1000 non-null   object
15  cogs                                   1000 non-null   float64
16  gross margin percentage                 1000 non-null   float64
17  gross income                           1000 non-null   float64
18  Rating                                 1000 non-null   float64
19  Day                                    1000 non-null   object
20  Date(dd-mm-yyyy)                       1000 non-null   object
21  Month                                  1000 non-null   object
dtypes: float64(7), int64(4), object(11)
memory usage: 172.0+ KB

```

6 SALES BY CITY

```

[109]: data.groupby(['City' , 'Total']).size().reset_index()

```

```

[109]:
   City      Total  0
0  Mandalay  18.6375  1
1  Mandalay  20.1075  1
2  Mandalay  26.7225  1

```

3	Mandalay	28.4235	1
4	Mandalay	30.9960	1
..
992	Yangon	926.9505	1
993	Yangon	931.0350	1
994	Yangon	932.3370	1
995	Yangon	951.8250	1
996	Yangon	1039.2900	1

[997 rows x 3 columns]

```
[110]: sales=data.groupby('City' )['Total'].sum().reset_index()
sales
```

```
[110]:      City      Total
0  Mandalay 106197.6720
1  Naypyitaw 110568.7065
2    Yangon 106200.3705
```

```
[111]: sales_graph= px.pie(sales , names='City' , values='Total' , title='TOTAL Sales_
↳by city' )
sales_graph.show()
```

6.1 CALCULATING CUSTOMER SATISFACTION INDEX

```
[112]: rating=data.groupby(['City' , 'Product line' , 'Rating']).size().
↳reset_index(name='Count of ratings occurred')
rating.head(50)
```

```
[112]:      City      Product line  Rating  Count of ratings occurred
0  Mandalay  Electronic accessories    4.0                1
1  Mandalay  Electronic accessories    4.2                1
2  Mandalay  Electronic accessories    4.3                1
3  Mandalay  Electronic accessories    4.8                1
4  Mandalay  Electronic accessories    4.9                1
5  Mandalay  Electronic accessories    5.1                1
6  Mandalay  Electronic accessories    5.3                1
7  Mandalay  Electronic accessories    5.5                1
8  Mandalay  Electronic accessories    5.6                1
9  Mandalay  Electronic accessories    5.7                1
10 Mandalay  Electronic accessories    5.8                1
11 Mandalay  Electronic accessories    6.0                3
12 Mandalay  Electronic accessories    6.1                3
13 Mandalay  Electronic accessories    6.2                2
14 Mandalay  Electronic accessories    6.3                1
15 Mandalay  Electronic accessories    6.5                3
16 Mandalay  Electronic accessories    6.6                1
```


17	Mandalay	Electronic accessories	6.7	2
18	Mandalay	Electronic accessories	6.8	1
19	Mandalay	Electronic accessories	7.0	1
20	Mandalay	Electronic accessories	7.1	1
21	Mandalay	Electronic accessories	7.3	1
22	Mandalay	Electronic accessories	7.6	3
23	Mandalay	Electronic accessories	7.7	1
24	Mandalay	Electronic accessories	7.8	2
25	Mandalay	Electronic accessories	7.9	1
26	Mandalay	Electronic accessories	8.0	2
27	Mandalay	Electronic accessories	8.1	2
28	Mandalay	Electronic accessories	8.2	1
29	Mandalay	Electronic accessories	8.6	2
30	Mandalay	Electronic accessories	8.7	1
31	Mandalay	Electronic accessories	8.8	1
32	Mandalay	Electronic accessories	8.9	2
33	Mandalay	Electronic accessories	9.0	2
34	Mandalay	Electronic accessories	9.4	1
35	Mandalay	Electronic accessories	9.5	1
36	Mandalay	Electronic accessories	9.8	1
37	Mandalay	Electronic accessories	9.9	1
38	Mandalay	Electronic accessories	10.0	1
39	Mandalay	Fashion accessories	4.1	2
40	Mandalay	Fashion accessories	4.2	1
41	Mandalay	Fashion accessories	4.3	1
42	Mandalay	Fashion accessories	4.4	1
43	Mandalay	Fashion accessories	4.5	1
44	Mandalay	Fashion accessories	4.6	2
45	Mandalay	Fashion accessories	4.7	1
46	Mandalay	Fashion accessories	4.8	2
47	Mandalay	Fashion accessories	4.9	1
48	Mandalay	Fashion accessories	5.0	2
49	Mandalay	Fashion accessories	5.1	1

```
[113]: max_rating=rating['Rating'].max() # maximum rating is 10
max_rating
```

```
[113]: np.float64(10.0)
```

```
[114]: rating['Count of ratings occurred'].max()
```

```
[114]: np.int64(5)
```

6.1.1 calculation of csi for each product line in the city

```
[115]: Q=rating.query("`Product line` == 'Electronic accessories' and City ==  
↳ 'Mandalay'")  
Q
```

```
[115]:
```

	City	Product line	Rating	Count of ratings occurred
0	Mandalay	Electronic accessories	4.0	1
1	Mandalay	Electronic accessories	4.2	1
2	Mandalay	Electronic accessories	4.3	1
3	Mandalay	Electronic accessories	4.8	1
4	Mandalay	Electronic accessories	4.9	1
5	Mandalay	Electronic accessories	5.1	1
6	Mandalay	Electronic accessories	5.3	1
7	Mandalay	Electronic accessories	5.5	1
8	Mandalay	Electronic accessories	5.6	1
9	Mandalay	Electronic accessories	5.7	1
10	Mandalay	Electronic accessories	5.8	1
11	Mandalay	Electronic accessories	6.0	3
12	Mandalay	Electronic accessories	6.1	3
13	Mandalay	Electronic accessories	6.2	2
14	Mandalay	Electronic accessories	6.3	1
15	Mandalay	Electronic accessories	6.5	3
16	Mandalay	Electronic accessories	6.6	1
17	Mandalay	Electronic accessories	6.7	2
18	Mandalay	Electronic accessories	6.8	1
19	Mandalay	Electronic accessories	7.0	1
20	Mandalay	Electronic accessories	7.1	1
21	Mandalay	Electronic accessories	7.3	1
22	Mandalay	Electronic accessories	7.6	3
23	Mandalay	Electronic accessories	7.7	1
24	Mandalay	Electronic accessories	7.8	2
25	Mandalay	Electronic accessories	7.9	1
26	Mandalay	Electronic accessories	8.0	2
27	Mandalay	Electronic accessories	8.1	2
28	Mandalay	Electronic accessories	8.2	1
29	Mandalay	Electronic accessories	8.6	2
30	Mandalay	Electronic accessories	8.7	1
31	Mandalay	Electronic accessories	8.8	1
32	Mandalay	Electronic accessories	8.9	2
33	Mandalay	Electronic accessories	9.0	2
34	Mandalay	Electronic accessories	9.4	1
35	Mandalay	Electronic accessories	9.5	1
36	Mandalay	Electronic accessories	9.8	1
37	Mandalay	Electronic accessories	9.9	1
38	Mandalay	Electronic accessories	10.0	1

```
[116]: Q.groupby('City')['Rating'].mean().reset_index()
```

```
[116]:      City      Rating
0  Mandalay  7.069231
```

6.1.2 formual of CSI $\text{sum of ratings} / \text{total number of ratings} * 100 / \text{Max rating}$

```
[117]: total_rating=Q['Count of ratings occured'].sum()
total_rating
```

```
[117]: np.int64(55)
```

```
[118]: CSI_electronics=(Q['Rating']*Q['Count of ratings occured']).sum()/total_rating/
↪max_rating*100
CSI_electronics
```

```
[118]: np.float64(71.16363636363636)
```

```
[119]: cities=rating['City'].unique()
cities
```

```
[119]: array(['Mandalay', 'Naypyitaw', 'Yangon'], dtype=object)
```

```
[120]: product_lines = rating["Product line"].unique()
cities=rating['City'].unique()

CSI_DATA= []

for city in cities:
    for line in product_lines:
        Q = rating.query("`Product line` == @line and `City` == @city")
        total_rating = Q["Count of ratings occured"].sum()
        weighted_sum = (Q["Rating"] * Q["Count of ratings occured"]).sum()
        max_rating = 10
        csi = (weighted_sum / total_rating) / max_rating * 100
        print(f"{city}: {line}: {csi:.2f}%")

        CSI_DATA.append({
            "City": city,
            "Product line": line,
            "CSI(%)":round(csi,2)})

CSI_DF = pd.DataFrame(CSI_DATA)
```

Mandalay: Electronic accessories: 71.16%

Mandalay: Fashion accessories: 67.23%

Mandalay: Food and beverages: 69.94%

Mandalay: Health and beauty: 71.00%

Mandalay: Home and lifestyle: 65.16%
 Mandalay: Sports and travel: 65.10%
 Naypyitaw: Electronic accessories: 67.47%
 Naypyitaw: Fashion accessories: 74.40%
 Naypyitaw: Food and beverages: 70.80%
 Naypyitaw: Health and beauty: 69.98%
 Naypyitaw: Home and lifestyle: 70.60%
 Naypyitaw: Sports and travel: 70.29%
 Yangon: Electronic accessories: 69.12%
 Yangon: Fashion accessories: 68.78%
 Yangon: Food and beverages: 72.53%
 Yangon: Health and beauty: 69.00%
 Yangon: Home and lifestyle: 69.31%
 Yangon: Sports and travel: 72.58%

```
[121]: CSI_DF
```

```
[121]:
```

	City	Product line	CSI(%)
0	Mandalay	Electronic accessories	71.16
1	Mandalay	Fashion accessories	67.23
2	Mandalay	Food and beverages	69.94
3	Mandalay	Health and beauty	71.00
4	Mandalay	Home and lifestyle	65.16
5	Mandalay	Sports and travel	65.10
6	Naypyitaw	Electronic accessories	67.47
7	Naypyitaw	Fashion accessories	74.40
8	Naypyitaw	Food and beverages	70.80
9	Naypyitaw	Health and beauty	69.98
10	Naypyitaw	Home and lifestyle	70.60
11	Naypyitaw	Sports and travel	70.29
12	Yangon	Electronic accessories	69.12
13	Yangon	Fashion accessories	68.78
14	Yangon	Food and beverages	72.53
15	Yangon	Health and beauty	69.00
16	Yangon	Home and lifestyle	69.31
17	Yangon	Sports and travel	72.58

```
[122]: m_graph=px.bar(CSI_DF, y='CSI(%)' , x='City' , title='CSI OF CITY'
↵, barmode='group', color='Product line')

m_graph.show()
```

7 SALES PREDICTION MODEL FOR EACH CITY ACCORDING TO PRODUCT LINE

```
[123]: sg_data=data.groupby(['City' , 'Product line' ])[ 'Total' ].sum().
        ↪reset_index(name='Total Revenue')
sg_data
```

```
[123]:
```

	City	Product line	Total Revenue
0	Mandalay	Electronic accessories	17051.4435
1	Mandalay	Fashion accessories	16413.3165
2	Mandalay	Food and beverages	15214.8885
3	Mandalay	Health and beauty	19980.6600
4	Mandalay	Home and lifestyle	17549.1645
5	Mandalay	Sports and travel	19988.1990
6	Naypyitaw	Electronic accessories	18968.9745
7	Naypyitaw	Fashion accessories	21560.0700
8	Naypyitaw	Food and beverages	23766.8550
9	Naypyitaw	Health and beauty	16615.3260
10	Naypyitaw	Home and lifestyle	13895.5530
11	Naypyitaw	Sports and travel	15761.9280
12	Yangon	Electronic accessories	18317.1135
13	Yangon	Fashion accessories	16332.5085
14	Yangon	Food and beverages	17163.1005
15	Yangon	Health and beauty	12597.7530
16	Yangon	Home and lifestyle	22417.1955
17	Yangon	Sports and travel	19372.6995

```
[124]: sg=px.bar(sg_data , x='Product line' ,y='Total Revenue' , facet_row='City',
        ↪color='Product line', height=1200,
        title='TOTAL REVENUE OF EACH PRODUCT LINE IN DIFFERENT CITIES' )

sg.update_xaxes(showticklabels=True)
sg.show()
```

```
[136]: Month_sales=data.groupby(['City', 'Month' , 'Product line' ])[ 'Total' ].sum().
        ↪reset_index()
Month_sales
```

```
[136]:
```

	City	Month	Product line	Total
0	Mandalay	February	Electronic accessories	6686.2530
1	Mandalay	February	Fashion accessories	6137.1135
2	Mandalay	February	Food and beverages	5554.8150
3	Mandalay	February	Health and beauty	5856.4275
4	Mandalay	February	Home and lifestyle	4659.8475
5	Mandalay	February	Sports and travel	5529.8145
6	Mandalay	January	Electronic accessories	6699.7770
7	Mandalay	January	Fashion accessories	6112.5960

8	Mandalay	January	Food and beverages	6609.2775
9	Mandalay	January	Health and beauty	6399.8865
10	Mandalay	January	Home and lifestyle	4586.4420
11	Mandalay	January	Sports and travel	6768.0795
12	Mandalay	March	Electronic accessories	3665.4135
13	Mandalay	March	Fashion accessories	4163.6070
14	Mandalay	March	Food and beverages	3050.7960
15	Mandalay	March	Health and beauty	7724.3460
16	Mandalay	March	Home and lifestyle	8302.8750
17	Mandalay	March	Sports and travel	7690.3050
18	Naypyitaw	February	Electronic accessories	5473.8810
19	Naypyitaw	February	Fashion accessories	7699.1145
20	Naypyitaw	February	Food and beverages	7391.3175
21	Naypyitaw	February	Health and beauty	5830.3455
22	Naypyitaw	February	Home and lifestyle	3002.9055
23	Naypyitaw	February	Sports and travel	3537.4185
24	Naypyitaw	January	Electronic accessories	5730.2385
25	Naypyitaw	January	Fashion accessories	6385.0290
26	Naypyitaw	January	Food and beverages	8315.0235
27	Naypyitaw	January	Health and beauty	6020.6895
28	Naypyitaw	January	Home and lifestyle	5594.7045
29	Naypyitaw	January	Sports and travel	8388.9960
30	Naypyitaw	March	Electronic accessories	7764.8550
31	Naypyitaw	March	Fashion accessories	7475.9265
32	Naypyitaw	March	Food and beverages	8060.5140
33	Naypyitaw	March	Health and beauty	4764.2910
34	Naypyitaw	March	Home and lifestyle	5297.9430
35	Naypyitaw	March	Sports and travel	3835.5135
36	Yangon	February	Electronic accessories	5202.7710
37	Yangon	February	Fashion accessories	5173.6335
38	Yangon	February	Food and beverages	7054.2255
39	Yangon	February	Health and beauty	2915.4825
40	Yangon	February	Home and lifestyle	4771.6305
41	Yangon	February	Sports and travel	4742.3775
42	Yangon	January	Electronic accessories	6401.2725
43	Yangon	January	Fashion accessories	6847.4910
44	Yangon	January	Food and beverages	4646.2290
45	Yangon	January	Health and beauty	3962.5950
46	Yangon	January	Home and lifestyle	10313.5935
47	Yangon	January	Sports and travel	6509.9475
48	Yangon	March	Electronic accessories	6713.0700
49	Yangon	March	Fashion accessories	4311.3840
50	Yangon	March	Food and beverages	5462.6460
51	Yangon	March	Health and beauty	5719.6755
52	Yangon	March	Home and lifestyle	7331.9715
53	Yangon	March	Sports and travel	8120.3745

```
[126]: Mg=px.bar(Month_sales, x="Month", y="Total", color="Product line",
↪,facet_row="City", barmode='group', category_orders={"Month":["January",
↪,'February' , 'March']}},
        title="MONTHLY SALES OF PRODUCT LINE" ,height=1200)

Mg.update_xaxes(showticklabels=True)
Mg.show()
```

```
[135]: Month_sales=data.groupby(['City','Month' , 'Product line' ])[ 'Total' ].sum()
Month_sales
```

```
[135]: City      Month      Product line      Total
Mandalay  February  Electronic accessories  6686.2530
          February  Fashion accessories     6137.1135
          February  Food and beverages     5554.8150
          February  Health and beauty      5856.4275
          February  Home and lifestyle     4659.8475
          February  Sports and travel      5529.8145
          January   Electronic accessories  6699.7770
          January   Fashion accessories     6112.5960
          January   Food and beverages     6609.2775
          January   Health and beauty      6399.8865
          January   Home and lifestyle     4586.4420
          January   Sports and travel      6768.0795
          March     Electronic accessories  3665.4135
          March     Fashion accessories     4163.6070
          March     Food and beverages     3050.7960
          March     Health and beauty      7724.3460
          March     Home and lifestyle     8302.8750
          March     Sports and travel      7690.3050
Naypyitaw February  Electronic accessories  5473.8810
          February  Fashion accessories     7699.1145
          February  Food and beverages     7391.3175
          February  Health and beauty      5830.3455
          February  Home and lifestyle     3002.9055
          February  Sports and travel      3537.4185
          January   Electronic accessories  5730.2385
          January   Fashion accessories     6385.0290
          January   Food and beverages     8315.0235
          January   Health and beauty      6020.6895
          January   Home and lifestyle     5594.7045
          January   Sports and travel      8388.9960
          March     Electronic accessories  7764.8550
          March     Fashion accessories     7475.9265
          March     Food and beverages     8060.5140
          March     Health and beauty      4764.2910
          March     Home and lifestyle     5297.9430
```

Yangon	February	Sports and travel	3835.5135
		Electronic accessories	5202.7710
		Fashion accessories	5173.6335
		Food and beverages	7054.2255
		Health and beauty	2915.4825
		Home and lifestyle	4771.6305
	January	Sports and travel	4742.3775
		Electronic accessories	6401.2725
		Fashion accessories	6847.4910
		Food and beverages	4646.2290
		Health and beauty	3962.5950
		Home and lifestyle	10313.5935
	March	Sports and travel	6509.9475
		Electronic accessories	6713.0700
		Fashion accessories	4311.3840
		Food and beverages	5462.6460
		Health and beauty	5719.6755
		Home and lifestyle	7331.9715
		Sports and travel	8120.3745

Name: Total, dtype: float64

```
[137]: Month_sales.groupby(['City', 'Month'])['Total'].sum()
```

```
[137]: City      Month
Mandalay  February    34424.2710
          January     37176.0585
          March       34597.3425
Naypyitaw February    32934.9825
          January     40434.6810
          March       37199.0430
Yangon    February    29860.1205
          January     38681.1285
          March       37659.1215
```

Name: Total, dtype: float64

```
[138]: sales_df=Month_sales.groupby(['City', 'Month'])['Total'].sum().
        ↪reset_index(name="Revenue")
sales_df
```

```
[138]:   City      Month      Revenue
0  Mandalay  February    34424.2710
1  Mandalay   January    37176.0585
2  Mandalay    March     34597.3425
3  Naypyitaw February    32934.9825
4  Naypyitaw   January    40434.6810
5  Naypyitaw    March     37199.0430
6    Yangon  February    29860.1205
```



```

7      Yangon    January  38681.1285
8      Yangon     March  37659.1215

```

```
[139]: sales_df['Month']=pd.Categorical(sales_df['Month'],categories=['January',
↳'February', 'March'], ordered =True)
```

```
[140]: final_sales=sales_df.groupby(['City', 'Month', 'Revenue'],observed=True).size().
↳reset_index()
final_sales
```

```
[140]:
```

	City	Month	Revenue	0
0	Mandalay	January	37176.0585	1
1	Mandalay	February	34424.2710	1
2	Mandalay	March	34597.3425	1
3	Naypyitaw	January	40434.6810	1
4	Naypyitaw	February	32934.9825	1
5	Naypyitaw	March	37199.0430	1
6	Yangon	January	38681.1285	1
7	Yangon	February	29860.1205	1
8	Yangon	March	37659.1215	1

```
[141]: final_sales=sales_df.groupby(['City', 'Month', 'Revenue'],observed=True).size().
↳reset_index().drop(columns=0)
final_sales
```

```
[141]:
```

	City	Month	Revenue
0	Mandalay	January	37176.0585
1	Mandalay	February	34424.2710
2	Mandalay	March	34597.3425
3	Naypyitaw	January	40434.6810
4	Naypyitaw	February	32934.9825
5	Naypyitaw	March	37199.0430
6	Yangon	January	38681.1285
7	Yangon	February	29860.1205
8	Yangon	March	37659.1215

```
[142]: sg2=px.line(final_sales,y='Revenue', x='Month',title='MONTHLY REVENUE OF_
↳CITIES',
            markers=True, height=600, color='City')
sg2.update_xaxes(showticklabels=True)
sg2.show()
```

8 REVENUE PREDICTION FOR APRIL MONTH

```
[143]: final_sales
```

```
[143]:
```

	City	Month	Revenue
0	Mandalay	January	37176.0585
1	Mandalay	February	34424.2710
2	Mandalay	March	34597.3425
3	Naypyitaw	January	40434.6810
4	Naypyitaw	February	32934.9825
5	Naypyitaw	March	37199.0430
6	Yangon	January	38681.1285
7	Yangon	February	29860.1205
8	Yangon	March	37659.1215

```
[174]: pvt_table=final_sales.pivot_table(index='City' , columns='Month' ,
    ↪values='Revenue' , observed=True)

pvt_table
```

```
[174]:
```

Month	January	February	March
City			
Mandalay	37176.0585	34424.2710	34597.3425
Naypyitaw	40434.6810	32934.9825	37199.0430
Yangon	38681.1285	29860.1205	37659.1215

```
[175]: X=pvt_table[['January' , 'February' , 'March']] # X=feature
y=pvt_table['March'] # y == target

model= LinearRegression()
model.fit(X,y)

pvt_table['April_Prediction']=model.predict(X)

sg_predict=pvt_table[['January' , 'February' , 'March' , 'April_Prediction']]
sg_predict
```

```
[175]:
```

Month	January	February	March	April_Prediction
City				
Mandalay	37176.0585	34424.2710	34597.3425	34597.3425
Naypyitaw	40434.6810	32934.9825	37199.0430	37199.0430
Yangon	38681.1285	29860.1205	37659.1215	37659.1215

```
[176]: pvt_table = pvt_table.reset_index()

pvt_table
```

```
[176]:
```

Month	City	January	February	March	April_Prediction
0	Mandalay	37176.0585	34424.2710	34597.3425	34597.3425
1	Naypyitaw	40434.6810	32934.9825	37199.0430	37199.0430
2	Yangon	38681.1285	29860.1205	37659.1215	37659.1215

```
[177]: pvt_table[['City', 'January', 'February', 'March']]
```

```
[177]:
```

	Month	City	January	February	March
0		Mandalay	37176.0585	34424.2710	34597.3425
1		Naypyitaw	40434.6810	32934.9825	37199.0430
2		Yangon	38681.1285	29860.1205	37659.1215

```
[178]: sg_long = pd.melt(
    pvt_table,
    id_vars=['City'],
    value_vars=['January', 'February', 'March', 'April_Prediction'],
    var_name='Month',
    value_name='Revenue'
)

sg_long
```

```
[178]:
```

	City	Month	Revenue
0	Mandalay	January	37176.0585
1	Naypyitaw	January	40434.6810
2	Yangon	January	38681.1285
3	Mandalay	February	34424.2710
4	Naypyitaw	February	32934.9825
5	Yangon	February	29860.1205
6	Mandalay	March	34597.3425
7	Naypyitaw	March	37199.0430
8	Yangon	March	37659.1215
9	Mandalay	April_Prediction	34597.3425
10	Naypyitaw	April_Prediction	37199.0430
11	Yangon	April_Prediction	37659.1215

```
[185]: sg_predict_graph=px.line(sg_long , x='Month' ,y='Revenue' ,color='City' ,
    ↪markers=True)
sg_predict_graph.show()
```

9 NOW I'LL BE PUTTING ALL OF HE GRAPHS IN THE FORM OF DASBOARD

9.0.1 installing pandoc to export it as PDF

```
[ ]:
```