#Author : Harsha Kumavat

GRIP @The Sparks Foundation Internship - JAN 2022

Data Science & Business Analytics Intern

Task 3: Exploratory Data Analysis - Retail.

Perform 'Exploratory Data Analysis' on dataset 'SampleSuperstore'. As a business manager, try to find out the weak areas where you can work to make more profit. What all business problems you can derive by exploring the data?

Importing Libraries

```
In [1]: #Import the necessary Libraries
    import pandas as pd
    import plotly.express as px
    from plotly.subplots import make_subplots
    import plotly.graph_objects as go
    import warnings
    warnings.filterwarnings('ignore')
```

Importing the dataset

```
In [2]: # Load the dataset
df = pd.read_csv('SampleSuperstore (1).csv')
```

In [3]: df.head() #view first 5 rows of the dataset

Out[3]:

	Ship Mode	Segment	Country	City	State	Postal Code	Region	Category	Sub-Category	Sales	Quantity	Discount	Profit
0	Second Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	Bookcases	261.9600	2	0.00	41.9136
1	Second Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	Chairs	731.9400	3	0.00	219.5820
2	Second Class	Corporate	United States	Los Angeles	California	90036	West	Office Supplies	Labels	14.6200	2	0.00	6.8714
3	Standard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Furniture	Tables	957.5775	5	0.45	-383.0310
4	Standard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Office Supplies	Storage	22,3680	2	0.20	2.5164

In [4]: df.tail() #view last 5 rows of the dataset

Out[4]:

	Ship Mode	Segment	Country	City	State	Postal Code	Region	Category	Sub-Category	Sales	Quantity	Discount	Profit
9989	Second Class	Consumer	United States	Miami	Florida	33180	South	Furniture	Furnishings	25.248	3	0.2	4.1028
9990	Standard Class	Consumer	United States	Costa Mesa	California	92627	West	Furniture	Furnishings	91.960	2	0.0	15.6332
9991	Standard Class	Consumer	United States	Costa Mesa	California	92627	West	Technology	Phones	258,576	2	0.2	19.3932
9992	Standard Class	Consumer	United States	Costa Mesa	California	92627	West	Office Supplies	Paper	29.600	4	0.0	13.3200
9993	Second Class	Consumer	United States	Westminster	California	92683	West	Office Supplies	Appliances	243.160	2	0.0	72.9480

Exploratory Data analysis

In [5]: df.shape #returns the no of rows and columns

Out[5]: (9994, 13)

```
In [6]: df.info()
                  #Basic summary about the data
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 9994 entries, 0 to 9993
        Data columns (total 13 columns):
         # Column
                         Non-Null Count Dtype
        --- -----
                         -----
         0 Ship Mode
                       9994 non-null object
         1
            Segment
                         9994 non-null object
         2
            Country
                         9994 non-null object
            City
                         9994 non-null object
         3
             State
                         9994 non-null
                                        object
             Postal Code 9994 non-null int64
                         9994 non-null object
         6
             Region
         7
                         9994 non-null object
            Category
            Sub-Category 9994 non-null object
         8
                         9994 non-null float64
         9 Sales
         10 Ouantity
                         9994 non-null int64
         11 Discount
                         9994 non-null float64
                         9994 non-null float64
         12 Profit
        dtypes: float64(3), int64(2), object(8)
        memory usage: 1015.1+ KB
In [7]: df.isnull().sum() #checking whether any null values are present
Out[7]: Ship Mode
        Segment
                       0
        Country
                       0
        City
        State
        Postal Code
        Region
        Category
        Sub-Category
        Sales
                       0
        Quantity
                       0
        Discount
        Profit
                       0
        dtype: int64
In [8]: df.duplicated().sum()
Out[8]: 17
In [9]: df.drop_duplicates(inplace=True)
In [10]: df.columns
Out[10]: Index(['Ship Mode', 'Segment', 'Country', 'City', 'State', 'Postal Code',
               'Region', 'Category', 'Sub-Category', 'Sales', 'Quantity', 'Discount',
               'Profit'],
              dtype='object')
```

```
In [11]: df.nunique() #gives the count of unique values present in the particular column
Out[11]: Ship Mode
                             4
          Segment
                             3
         Country
                             1
         City
                           531
         State
                            49
         Postal Code
                           631
         Region
                             4
         Category
                             3
         Sub-Category
                            17
         Sales
                          5825
         Quantity
                            14
         Discount
                            12
         Profit
                          7287
         dtype: int64
In [12]: df.drop(columns='Postal Code',axis=1,inplace=True)
In [13]: df.describe() #Statistical summary of data
Out[13]:
                                                        Profit
                       Sales
                                Quantity
                                          Discount
                 9977.000000 9977.000000 9977.000000
                                                   9977.00000
          count
                               3.790719
                                          0.156278
                                                     28.69013
                   230.148902
           mean
                   623.721409
                               2.226657
                                          0.206455
                                                    234.45784
             std
                    0.444000
                               1.000000
                                          0.000000 -6599.97800
            min
            25%
                   17.300000
                               2.000000
                                          0.000000
                                                      1.72620
            50%
                   54.816000
                               3.000000
                                          0.200000
                                                      8.67100
                   209.970000
                               5.000000
                                          0.200000
                                                     29.37200
            max 22638.480000
                              14.000000
                                          0.800000
                                                   8399.97600
In [14]: df['Ship Mode'].value_counts().to_frame()
Out[14]:
                        Ship Mode
                             5955
           Standard Class
            Second Class
                             1943
                             1537
              First Class
               Same Day
                             542
In [15]: df['Segment'].value_counts().to_frame()
Out[15]:
                      Segment
            Consumer
                         5183
            Corporate
                         3015
           Home Office
                         1779
In [16]: df['Country'].value_counts().to_frame()
Out[16]:
                       Country
          United States
                         9977
```

 West
 3193

 East
 2845

 Central
 2319

 South
 1620

```
In [18]: a = pd.DataFrame(df.groupby('Ship Mode')['Sales'].sum().sort_values(ascending=False))
a.reset_index(inplace=True)
a.columns=['Ship Mode','Sales']

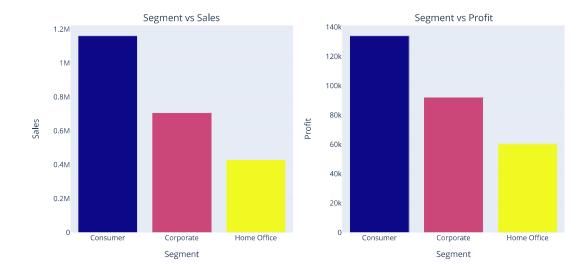
b = pd.DataFrame(df.groupby('Ship Mode')['Profit'].sum().sort_values(ascending=False))
b.reset_index(inplace=True)
b.columns=['Ship Mode','Profit']

fig = make_subplots(rows=1,cols=2,subplot_titles=("Ship Mode vs Sales","Ship Mode vs Profit", ))
fig.add_trace(go.Bar(x=a'|Ship Mode'), y=a'|Sales'],marker=dict(color=[1,2,3,4,5])),1, 1)
fig.add_trace(go.Bar(x=a'|Ship Mode'), y=b'|Profit'],marker=dict(color=[1,2,3,4,5])),1, 2)

fig.update_xaxes(title_text="Ship Mode", row=1, col=1)
fig.update_xaxes(title_text="Ship Mode", row=1, col=2)
fig.update_yaxes(title_text="Ship Mode", row=1, col=2)
fig.update_yaxes(title_text="Ship Mode", row=1, col=2)
fig.update_layout(showlegend=False)
fig.show()
```



```
In [19]: a = pd.DataFrame(df.groupby('Segment')['Profit'].sum().sort_values(ascending=False))
         a.reset_index(inplace=True)
         a.columns=['Segment','Profit']
         b = pd.DataFrame(df.groupby('Segment')['Sales'].sum().sort_values(ascending=False))
         b.reset_index(inplace=True)
         b.columns=['Segment','Sales']
         fig = make_subplots(rows=1,cols=2,subplot_titles=("Segment vs Sales","Segment vs Profit"))
         fig.add_trace(go.Bar(x=b['Segment'],y=b['Sales'],marker=dict(color=[1,2,3])),1, 1)
         fig.add_trace(go.Bar(x=a['Segment'], y=a['Profit'], marker=dict(color=[1,2,3])),1, 2)
         fig.update_layout(showlegend=False)
         fig.update_xaxes(title_text="Segment", row=1, col=1)
         fig.update_xaxes(title_text="Segment", row=1, col=2)
         fig.update_yaxes(title_text="Sales", row=1, col=1)
         fig.update_yaxes(title_text="Profit",row=1, col=2)
         fig.update_layout(showlegend=False)
         fig.show()
```



```
In [20]: df.groupby('Category')['Sub-Category'].value_counts().to_frame()
Out[20]:
```

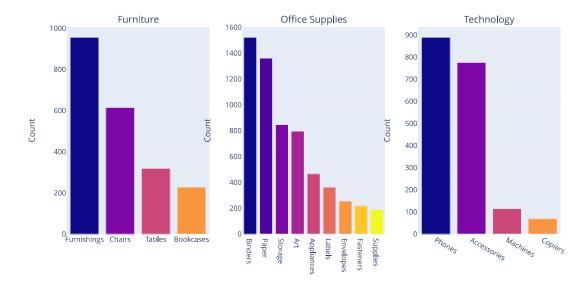
Sub-Category

Category	Sub-Category	
Furniture	Furnishings	956
	Chairs	615
	Tables	319
	Bookcases	228
Office Supplies	Binders	1522
	Paper	1359
	Storage	846
	Art	795
	Appliances	466
	Labels	363
	Envelopes	254
	Fasteners	217
	Supplies	190
Technology	Phones	889
	Accessories	775
	Machines	115
	Copiers	68

```
In [21]: Furniture = pd.DataFrame(df[df['Category'] == 'Furniture']['Sub-Category'].value_counts())
Furniture.reset_index(inplace=True)
Furniture.columns = ['Furniture', 'Count']
Office_Supplies = pd.DataFrame(df[df['Category'] == 'Office Supplies']['Sub-Category'].value_counts())
Office_Supplies.reset_index(inplace=True)
Office_Supplies.columns = ['Office_Supplies', 'Count']
Technology = pd.DataFrame(df[df['Category'] == 'Technology']['Sub-Category'].value_counts())
Technology.reset_index(inplace=True)
Technology.columns = ['Technology', 'Count']
```

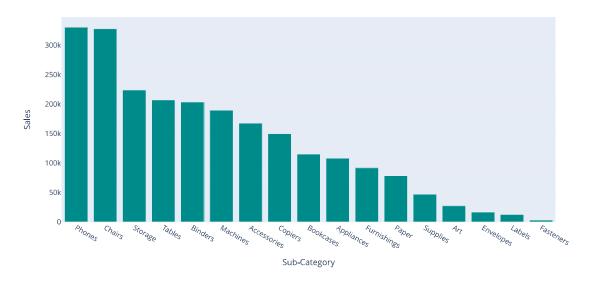
```
In [22]: fig = make_subplots(rows=1,cols=3,subplot_titles=("Furniture","Office Supplies", "Technology"))
    fig.add_trace(go.Bar(x=Furniture['Furniture'], y=Furniture['Count'],marker=dict(color=[1,2,3,4,5])),1, 1)
    fig.add_trace(go.Bar(x=Office_Supplies['Office_Supplies['Count'],marker=dict(color=[1,2,3,4,5,6,7,8,9])),1, 2)
    fig.add_trace(go.Bar(x=Technology['Technology'], y=Technology['Count'],marker=dict(color=[1,2,3,4,5])),1, 3)

    fig.update_yaxes(title_text="Count",row=1, col=2)
    fig.update_yaxes(title_text="Count",row=1, col=1)
    fig.update_layout(showlegend=False)
    fig.show()
```



```
In [23]: a = pd.DataFrame(df.groupby('Sub-Category')['Sales'].sum().sort_values(ascending=False))
a.reset_index(inplace=True)
a.columns=['Sub-Category','Sales']
fig = px.bar(a,y=a['Sales'],x=a['Sub-Category'],title='Sub-Category vs Sales',color_discrete_sequence=['DarkCyan'])
fig.show()
```

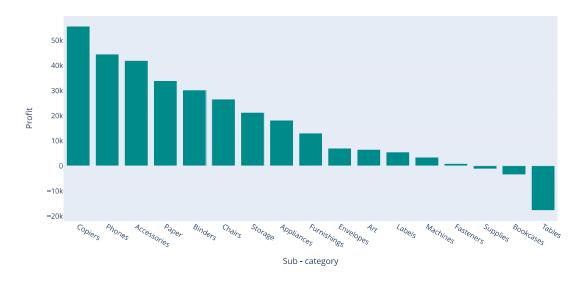
Sub-Category vs Sales



```
data = ['Sales','Quantity','Profit','Discount','State','Category','Sub-Category','Segment']
          data=df[data]
          data=data.sort_values(by='Profit',ascending=False)
          df1 = pd.pivot_table(data,index=['Category','Sub-Category'])
          df1
Out[24]:
                                      Discount
                                                    Profit Quantity
                                                                         Sales
                Category Sub-Category
                Furniture
                            Bookcases 0.211140 -15.230509 3.807018
                                                                    503.859633
                               Chairs 0.170244
                                                43.198582 3.822764
                                                                     532,971969
                           Furnishings 0.138494
                                                13.653476 3.723849
                                                                     95.902745
                                      0.261285
                                                -55.565771
                                                          3.890282
                                                                     648.794771
           Office Supplies
                           Appliances
                                      0.166524
                                                38.922758
                                                          3.710300
                                                                    230.755710
                                  Art
                                      0.074969
                                                 8.207059
                                                          3.768553
                                                                     34.096896
                                      0.372011
                                                19.860710 3.923127
                                                                     133.645972
                              Binders
                                                                     64.867724
                                      0.080315
                                                27.418019
                                                          3.566929
                                      0.082028
                                                 4.375660
                                                          4.211982
                                                                     13.936774
                             Fasteners
                               Labels
                                      0.068871
                                                15.224193 3.845730
                                                                     34.283504
                                Paper
                                      0.074908
                                                24.977365 3.785136
                                                                     57.560075
                              Storage
                                      0.074704
                                                25.152277 3.732861
                                                                    264.590553
                              Supplies
                                      0.076842
                                                 -6.258418 3.405263
              Technology Accessories
                                      0.078452
                                                54.111788 3.840000
                                                                    215,974604
                                      0.161765 817.909190 3.441176 2198.941618
                             Machines
                                      0.306087
                                                29.432669 3.826087 1645.553313
                              Phones 0.154556 50.073938 3.699663 371.211534
In [25]: data.pivot_table(values='Profit',index='Segment',columns='Discount',aggfunc='median')
Out[25]:
              Discount
                           0.00
                                  0.10
                                            0.15 0.20
                                                                                        0.45
                                                                                                 0.50
                                                                                                          0.60
                                                                                                                 0.70
                                                                                                                          0.80
              Segment
             Consumer 16.14600 62.0390
                                         5,12940 6,0433 -26,0562 -30,0980 -47,21360 -135,68145 -175,2600 -14,1323 -8,7276 -13,7175
             Corporate 16.35285 46.7908 26.27735 6.7068 -28.2240 -59.0606 -87.27495 -255.58750 -120.5130 -10.4196 -8.9796 -16.7130
           Home Office 15.45460 37.2300 16.79860 7.2576 -18.2220 -57.3234 -49.71900 -175.14690 -237.8425 -14.2290 -9.7608 -14.0328
```

```
In [26]: a = pd.DataFrame(df.groupby('Sub-Category')['Profit'].sum().sort_values(ascending=False))
a.reset_index(inplace=True)
a.columns=['Sub - category', 'Profit']
fig = px.bar(a,y=a['Profit'],x=a['Sub - category'],title='Sub-Category vs Profit',color_discrete_sequence=['DarkCyan'])
fig.show()
```

Sub-Category vs Profit



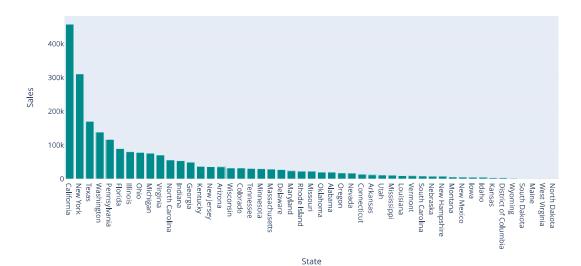
```
In [27]: a = pd.DataFrame(df.groupby('Region')['Profit'].sum().sort_values(ascending=False))
a.reset_index(inplace=True)
a.columns=['Region','Profit']
fig = px.bar(a,y=a['Profit'],x=a['Region'],title='Region vs Profit',color_discrete_sequence=['DarkCyan'],width=600,height=500)
fig.show()
```

Region vs Profit



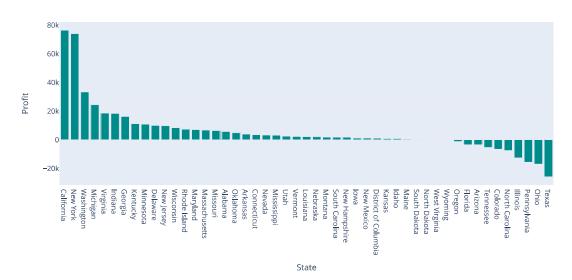
```
In [28]: a = pd.DataFrame(df.groupby('State')['Sales'].sum().sort_values(ascending=False))
    a.reset_index(inplace=True)
    a.columns=['State', 'Sales']
    fig = px.bar(a,y=a['Sales'],x=a['State'],title='State vs Sales',color_discrete_sequence=['DarkCyan'])
    fig.show()
```

State vs Sales



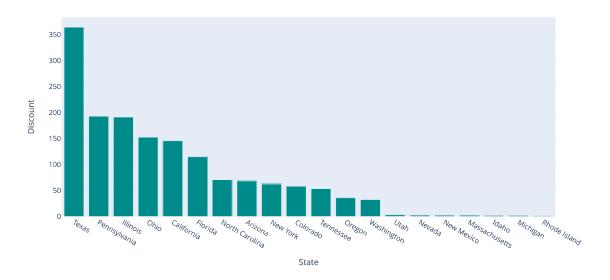
```
In [29]: a = pd.DataFrame(df.groupby('State')['Profit'].sum().sort_values(ascending=False))
a.reset_index(inplace=True)
a.columns=['State', 'Profit']
fig = px.bar(a,y=a['Profit'],x=a['State'],title='State vs Profit',color_discrete_sequence=['DarkCyan'])
fig.show()
```

State vs Profit



```
In [30]: a = pd.DataFrame(df.groupby('State')['Discount'].sum().sort_values(ascending=False)).head(20)
a.reset_index(inplace=True)
a.columns=['State', 'Discount']
fig = px.bar(a,y=a['Discount'],x=a['State'],title='State vs Discount',color_discrete_sequence=['darkcyan'])
fig.show()
```

State vs Discount



```
In [31]: a = pd.DataFrame(df.groupby('City')['Profit'].sum().sort_values(ascending=False).head(5))
a.reset_index(inplace=True)
a.columns=['City', 'Profit']
b = pd.DataFrame(df.groupby('City')['Profit'].sum().sort_values(ascending=False).tail(5))
b.reset_index(inplace=True)
b.columns=['City', 'Profit']

fig = make_subplots(rows=1,cols=2,subplot_titles=("Top 5 cities with max profit","Top 5 cities with min profit"))
fig.add_trace(go.Bar(x=a['City'],y=a['Profit'],marker=dict(color=[1,2,3,4,5])),1, 1)
fig.add_trace(go.Bar(x=b['City'],y=b['Profit'],marker=dict(color=[1,2,3,4,5])),1, 2)
fig.update_layout(showlegend=False)

fig.update_xaxes(title_text="Segment", row=1, col=2)
fig.update_vaxes(title_text="Profit", row=1, col=2)
fig.update_vaxes(title_text="Profit", row=1, col=2)
fig.update_layout(showlegend=False)
```



Conclusion

Problem Statement: Find out weak areas where you can work to make profit and what all business problem can be derived by exploring data.

- · Standard Class in ShipMode has recorded the highest profit and Same Day has recorded the lowest profit.
- There are 3 segments selling products they are Consumer, Corporate & Home Office where Consumer segment has recorded maximum profit followed by Corporate whereas Home Offices recorded minimum profit.
- In United States the products are sold where West region has recorded maximum profit followed by East and lowest being recorded in Central region.
- Top 5 most sold products Sub-Category wise are Phones, Chairs, Storage, Tables & Binders.
- Top 5 least sold products Sub-Category wise are Fasteners, Labels, Envelopes, Art & Supplies.
- When the discount given on a product is beyond 20% then company is getting a loss instead of gainning profit.
- Maximum profit is gained by Copiers, Phones, Accessories ,Paper, Binders whereas Tables has recorded maximim loss followed by Bookcases & Supplies. Hence discount given on these products can be reduced to increase profit.
- Maximum Sales are from states California, New York & Minimum sales are from North Dakota, West Virginia.
- State California & New Yok has recorded the maximum profit whereas Texas, Ohio, Pennsylvania in these states products has occured loss. So discount given in these states can be reduced to increase profit.
- As maximum sales are in states California, NewYork so sales can be increased in these areas to gain profit and In technology category company is getting benefitted so increase in sales of these category can increase profit.