

▼ P VAISHNAVI

DATA SCIENCE AND BUSINESS ANALYTICS

INTERN @ THE SPARKS FOUNDATION

DATASET : SAMPLESUPERSTORE.CSV (<https://bit.ly/3i4rbWl>)

EXPLORATORY DATA ANALYSIS - RETAIL

Task-1:

1. Perform 'Exploratory Data Analysis' on dataset 'SampleSuperstore'.
2. As a business manager, try to find out the weak areas where you can work to make more profit.

```
#importing the libraries
```

```
import numpy as np
import pandas as pd
import seaborn as sb
import matplotlib.pyplot as plt
```

```
from google.colab import files
uploaded = files.upload()
```

No file chosen

Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.

Saving SampleSuperstore.csv to SampleSuperstore.csv

```
#loading the dataset
```

```
data = pd.read_csv("SampleSuperstore.csv")
data
```

	Ship Mode	Segment	Country	City	State	Postal Code	Region	Category	Ca
0	Second Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	Boo
1	Second Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	
2	Second Class	Corporate	United States	Los Angeles	California	90036	West	Office Supplies	
3	Standard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Furniture	
4	Standard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Office Supplies	\$
...	
9989	Second Class	Consumer	United States	Miami	Florida	33180	South	Furniture	Furn
9990	Standard Class	Consumer	United States	Costa Mesa	California	92627	West	Furniture	Furn

```
data.head(5)
```

	Ship Mode	Segment	Country	City	State	Postal Code	Region	Category	Sub-Category
0	Second Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	Bookcases
1	Second Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	Chairs
2	Second Class	Corporate	United States	Los Angeles	California	90036	West	Office Supplies	Labels
3	Standard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Furniture	Tables
4	Standard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Office Supplies	Storage

```
data.tail(5)
```

	Ship Mode	Segment	Country	City	State	Postal Code	Region	Category	Ca
9989	Second Class	Consumer	United States	Miami	Florida	33180	South	Furniture	Furn
9990	Standard Class	Consumer	United States	Costa Mesa	California	92627	West	Furniture	Furn

```
data.info()
```

```
<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
3   City                 9994 non-null   object
4   State                9994 non-null   object
5   Postal Code          9994 non-null   int64
6   Region              9994 non-null   object
7   Category             9994 non-null   object
8   Sub-Category        9994 non-null   object
9   Sales               9994 non-null   float64
10  Quantity            9994 non-null   int64
11  Discount            9994 non-null   float64
12  Profit              9994 non-null   float64
dtypes: float64(3), int64(2), object(8)
memory usage: 1015.1+ KB
```

```
data.describe()
```

	Postal Code	Sales	Quantity	Discount	Profit
count	9994.000000	9994.000000	9994.000000	9994.000000	9994.000000
mean	55190.379428	229.858001	3.789574	0.156203	28.656896
std	32063.693350	623.245101	2.225110	0.206452	234.260108
min	1040.000000	0.444000	1.000000	0.000000	-6599.978000
25%	23223.000000	17.280000	2.000000	0.000000	1.728750
50%	56430.500000	54.490000	3.000000	0.200000	8.666500
75%	90008.000000	209.940000	5.000000	0.200000	29.364000
max	99301.000000	22638.480000	14.000000	0.800000	8399.976000

```
#detecting missing values in the dataset
data.isnull().sum()
```

```
Ship Mode      0
Segment        0
Country        0
City           0
State          0
Postal Code    0
Region         0
Category       0
Sub-Category   0
Sales          0
Quantity       0
Discount       0
Profit         0
dtype: int64
```

```
data.isna().sum()
```

```
Ship Mode      0
Segment        0
Country        0
City           0
State          0
Postal Code    0
Region         0
Category       0
Sub-Category   0
Sales          0
Quantity       0
Discount       0
Profit         0
dtype: int64
```

```
sales_data = data.groupby('Category', as_index=False)['Sales'].sum()
subcat_data = data.groupby(['Category', 'Sub-Category'])['Sales'].sum()
subcat_data['Sales']=map(int,subcat_data)
sales_data
```

	Category	Sales
0	Furniture	741999.7953
1	Office Supplies	719047.0320
2	Technology	836154.0330

```
data.columns
```

```
Index(['Ship Mode', 'Segment', 'Country', 'City', 'State', 'Region',
      'Category', 'Sub-Category', 'Sales', 'Quantity', 'Discount', 'Profit'],
      dtype='object')
```

```
data.index
```

```
RangeIndex(start=0, stop=9994, step=1)
```

```
data.nunique()
```

```

Ship Mode      4
Segment        3
Country         1
City           531
State          49
Region         4
Category        3
Sub-Category   17
Sales          5825
Quantity       14
Discount       12
Profit         7287
dtype: int64

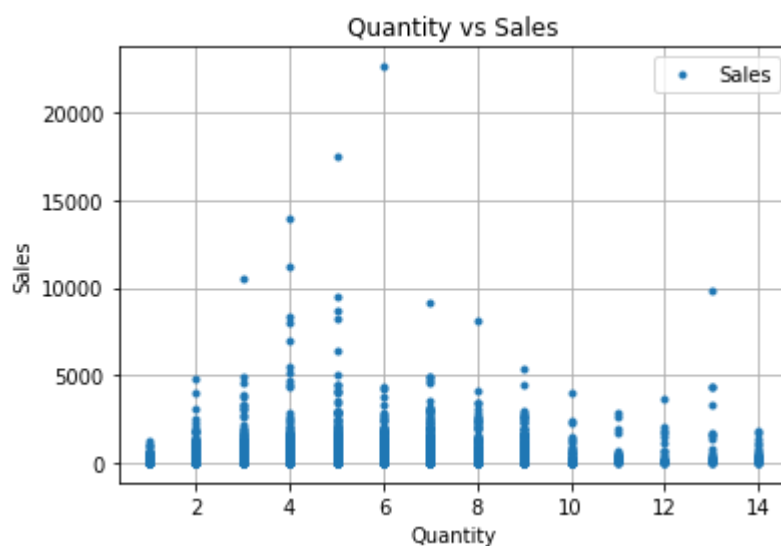
```

▼ Exploratory Data Analysis

```

data.plot(x='Quantity',y='Sales',style='.')
plt.title('Quantity vs Sales')
plt.xlabel('Quantity')
plt.ylabel('Sales')
plt.grid()
plt.show()

```

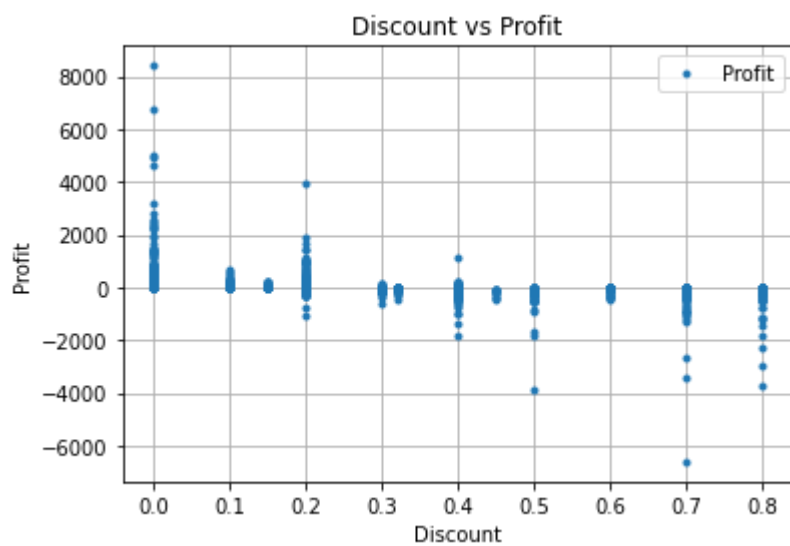


```

data.plot(x='Discount',y='Profit',style='.')
plt.title('Discount vs Profit')
plt.xlabel('Discount')
plt.ylabel('Profit')
plt.grid()

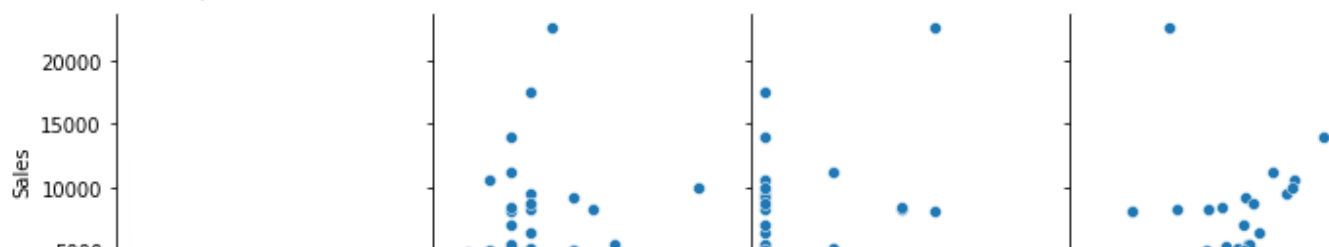
```

```
plt.show()
```



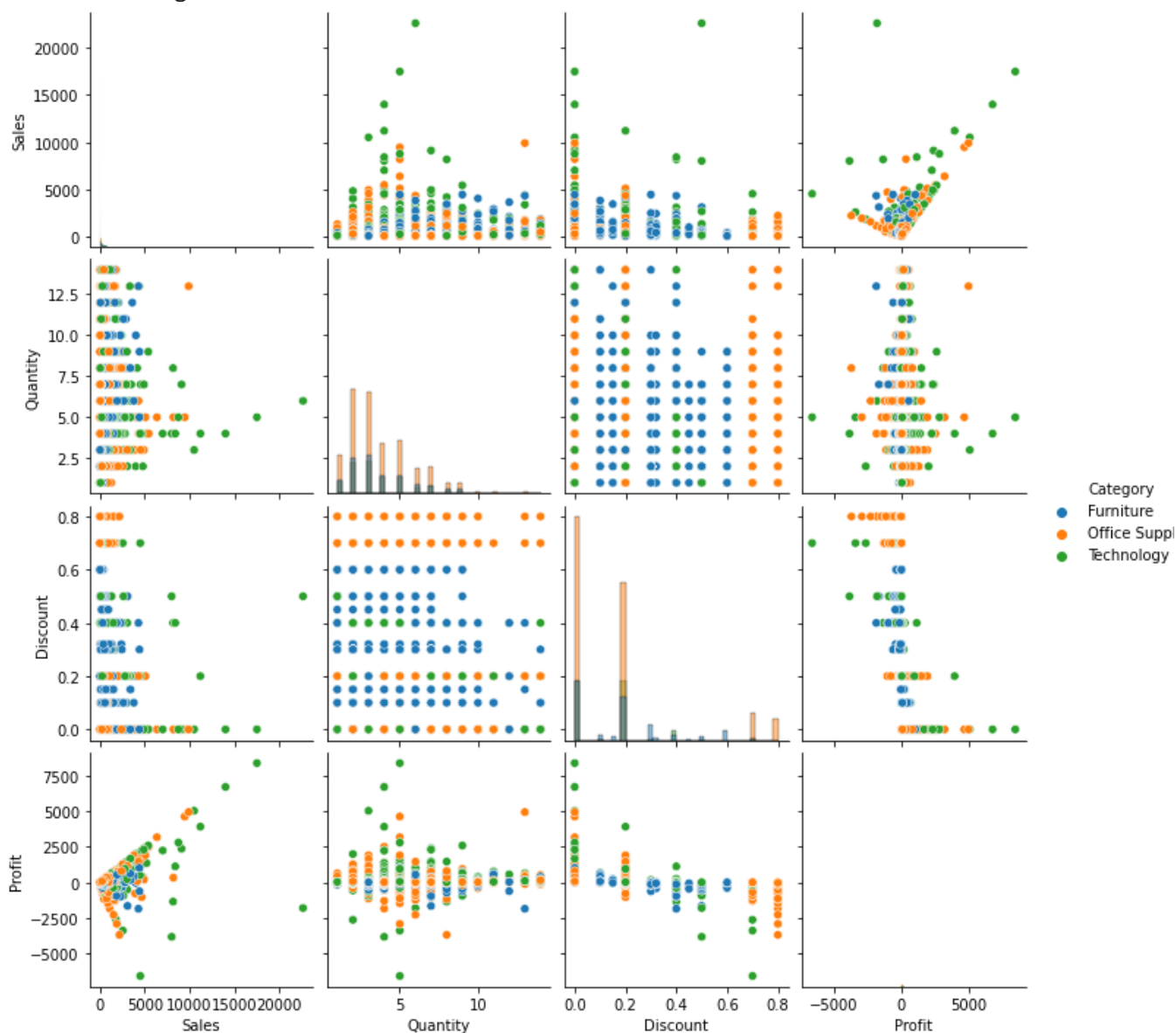
```
sb.pairplot(data)
```

```
<seaborn.axisgrid.PairGrid at 0x7fde9e233050>
```



```
sb.pairplot(data,hue='Category',diag_kind='hist')
```

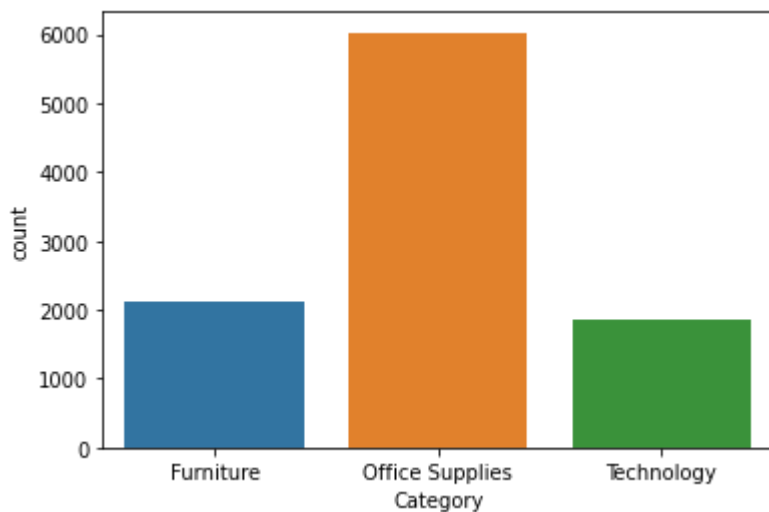
```
<seaborn.axisgrid.PairGrid at 0x7fde90a797d0>
```



```
data['Category'].value_counts()
```

```
sb.countplot(x=data['Category'])
```

<matplotlib.axes._subplots.AxesSubplot at 0x7fde90382450>



```
data.corr()
```

	Sales	Quantity	Discount	Profit
Sales	1.000000	0.200795	-0.028190	0.479064
Quantity	0.200795	1.000000	0.008623	0.066253
Discount	-0.028190	0.008623	1.000000	-0.219487
Profit	0.479064	0.066253	-0.219487	1.000000

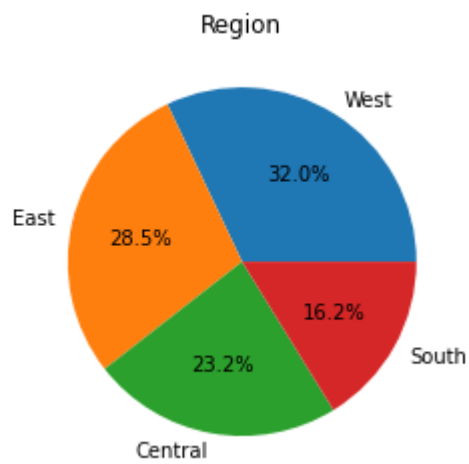
```
sb.heatmap(data.corr(), annot=True)
```

<matplotlib.axes._subplots.AxesSubplot at 0x7fde8d752810>

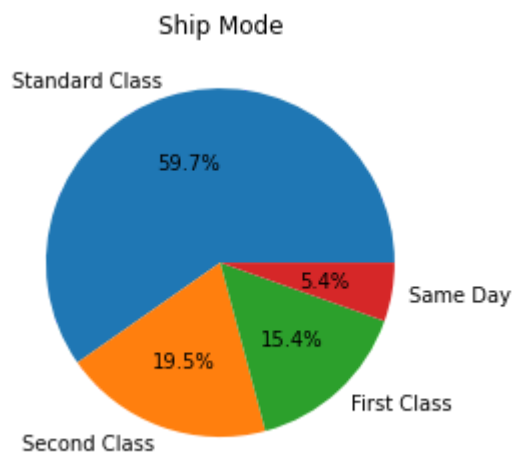


```
plt.title('Region')
```

```
plt.pie(data['Region'].value_counts(), labels=data['Region'].value_counts().index, autopct='%1.1%', show=True)
```

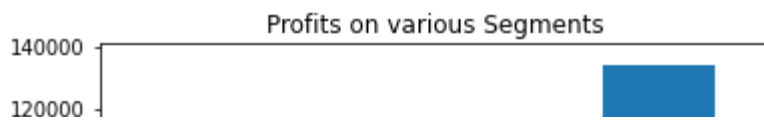



```
plt.title('Ship Mode')
plt.pie(data['Ship Mode'].value_counts(),labels=data['Ship Mode'].value_counts().index,autopc
plt.show()
```



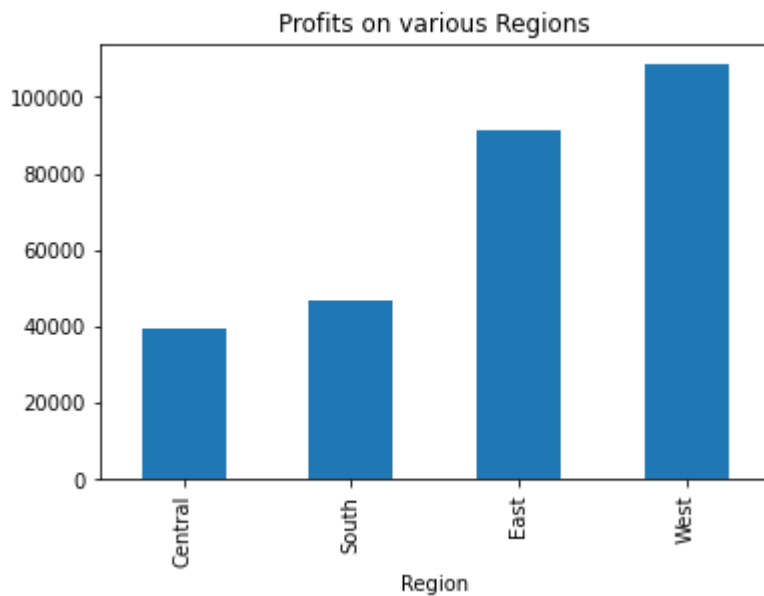
```
data.groupby('Segment')['Profit'].sum().sort_values().plot.bar()
plt.title("Profits on various Segments")
```

```
Text(0.5, 1.0, 'Profits on various Segments')
```



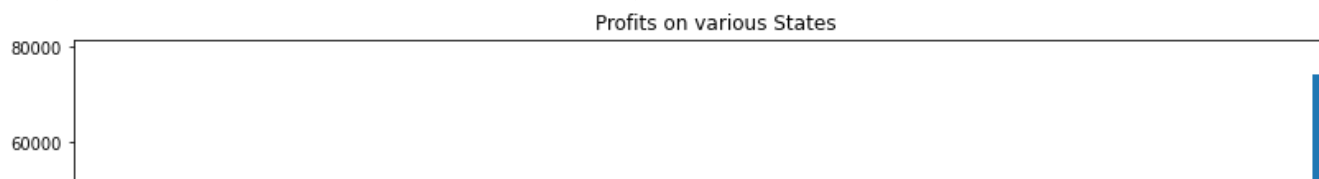
```
data.groupby('Region')['Profit'].sum().sort_values().plot.bar()  
plt.title("Profits on various Regions")
```

```
Text(0.5, 1.0, 'Profits on various Regions')
```



```
plt.figure(figsize=(14,6))  
data.groupby('State')['Profit'].sum().sort_values().plot.bar()  
plt.title("Profits on various States")
```

```
Text(0.5, 1.0, 'Profits on various States')
```



Statewise Deal Analysis

```
data['Country'].value_counts()
```

```
United States    9994
Name: Country, dtype: int64
```



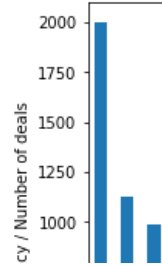
```
data1 = data['State'].value_counts()
data1.head(10)
```

```
California      2001
New York        1128
Texas           985
Pennsylvania    587
Washington      506
Illinois        492
Ohio            469
Florida         383
Michigan        255
North Carolina  249
Name: State, dtype: int64
```

```
data1.plot(kind='bar',figsize=(15,5))
plt.ylabel('Frequency / Number of deals')
plt.xlabel('States')
```

```
plt.title('State Wise Dealings', fontsize = 20)
plt.show()
```

State Wise Dealings



Here is top 3 state where deals are Highest.

Califonia

New York

Texas

Wyoming: Lowest Number of deal

```
data['State'].value_counts().mean()
```

203.9591836734694

Dis