

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

GATHERING THE DATASET

```
df=pd.read_csv("/content/SampleSuperstore.csv")
```

UNDERSTANDING THE DATASET

```
df.head()
```

	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	Laboratory Equipment
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

```
df.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

```
df.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
```

```
df.tail()
```

	Ship Mode	Segment	Country	City	State	Postal Code	Region	Category	Sub-Category	Sales	Quantity	Discount
9989	Second Class	Consumer	United States	Miami	Florida	33180	South	Furniture	Furnishings	25.248	3	0.2
9990	Standard Class	Consumer	United States	Costa Mesa	California	92627	West	Furniture	Furnishings	91.960	2	0.0

```
df["Category"].unique()

array(['Furniture', 'Office Supplies', 'Technology'], dtype=object)
```

```
df["Category"].value_counts()

Office Supplies    6026
Furniture          2121
Technology         1847
Name: Category, dtype: int64
```

DATA CLEANING

```
df.drop_duplicates(inplace=True)
```

df

	Ship Mode	Segment	Country	City	State	Postal Code	Region	Category	Sub-Category	Sales	Quantity	Discount
0	Second Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	Bookcases	261.9600	2	0.00
1	Second Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	Chairs	731.9400	3	0.00
2	Second Class	Corporate	United States	Los Angeles	California	90036	West	Office Supplies	Labels	14.6200	2	0.00
3	Standard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Furniture	Tables	957.5775	5	0.45
4	Standard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Office Supplies	Storage	22.3680	2	0.20
...
9989	Second Class	Consumer	United States	Miami	Florida	33180	South	Furniture	Furnishings	25.2480	3	0.20
9990	Standard Class	Consumer	United States	Costa Mesa	California	92627	West	Furniture	Furnishings	91.9600	2	0.00
9991	Standard Class	Consumer	United States	Costa Mesa	California	92627	West	Technology	Phones	258.5760	2	0.20
9992	Standard Class	Consumer	United States	Costa Mesa	California	92627	West	Office Supplies	Paper	29.6000	4	0.00
9993	Second Class	Consumer	United States	Westminster	California	92683	West	Office Supplies	Appliances	243.1600	2	0.00

9977 rows × 13 columns

```
df.drop(columns="Postal Code")
```

	Ship Mode	Segment	Country	City	State	Region	Category	Sub-Category	Sales	Quantity	Discount	Profit
0	Second Class	Consumer	United States	Henderson	Kentucky	South	Furniture	Bookcases	261.9600	2	0.00	41.9136
1	Second Class	Consumer	United States	Henderson	Kentucky	South	Furniture	Chairs	731.9400	3	0.00	219.5820

```
df.drop(columns="Postal Code", inplace=True)
```

```
df.head()
```

	Ship Mode	Segment	Country	City	State	Region	Category	Sub-Category	Sales	Quantity	Discount	Profit
0	Second Class	Consumer	United States	Henderson	Kentucky	South	Furniture	Bookcases	261.9600	2	0.00	41.9136
1	Second Class	Consumer	United States	Henderson	Kentucky	South	Furniture	Chairs	731.9400	3	0.00	219.5820
2	Second Class	Corporate	United States	Los Angeles	California	West	Office Supplies	Labels	14.6200	2	0.00	6.8714

```
print(df["Ship Mode"].unique())
print(df["Segment"].unique())
print(df["Country"].unique())
print(df["Category"].unique())
print(df["City"].unique())
print(df["State"].unique())
print(df["Region"].unique())
print(df["Sub-Category"].unique())
print(df["Sales"].unique())
print(df["Quantity"].unique())
print(df["Discount"].unique())
print(df["Profit"].unique())

['Second Class' 'Standard Class' 'First Class' 'Same Day']
['Consumer' 'Corporate' 'Home Office']
['United States']
['Furniture' 'Office Supplies' 'Technology']
['Henderson' 'Los Angeles' 'Fort Lauderdale' 'Concord' 'Seattle'
'Fort Worth' 'Madison' 'West Jordan' 'San Francisco' 'Fremont'
'Philadelphia' 'Orem' 'Houston' 'Richardson' 'Naperville' 'Melbourne'
'Eagan' 'Westland' 'Dover' 'New Albany' 'New York City' 'Troy' 'Chicago'
'Gilbert' 'Springfield' 'Jackson' 'Memphis' 'Decatur' 'Durham' 'Columbia'
'Rochester' 'Minneapolis' 'Portland' 'Saint Paul' 'Aurora' 'Charlotte'
'Orland Park' 'Urbandale' 'Columbus' 'Bristol' 'Wilmington' 'Bloomington'
'Phoenix' 'Roseville' 'Independence' 'Pasadena' 'Newark' 'Franklin'
'Scottsdale' 'San Jose' 'Edmond' 'Carlsbad' 'San Antonio' 'Monroe'
'Fairfield' 'Grand Prairie' 'Redlands' 'Hamilton' 'Westfield' 'Akron'
'Denver' 'Dallas' 'Whittier' 'Saginaw' 'Medina' 'Dublin' 'Detroit'
'Tampa' 'Santa Clara' 'Lakeville' 'San Diego' 'Brentwood' 'Chapel Hill'
'Morristown' 'Cincinnati' 'Inglewood' 'Tamarac' 'Colorado Springs'
'Belleville' 'Taylor' 'Lakewood' 'Arlington' 'Arvada' 'Hackensack'
'Saint Petersburg' 'Long Beach' 'Hesperia' 'Murfreesboro' 'Layton'
'Austin' 'Lowell' 'Manchester' 'Harlingen' 'Tucson' 'Quincy'
'Pembroke Pines' 'Des Moines' 'Peoria' 'Las Vegas' 'Warwick' 'Miami'
'Huntington Beach' 'Richmond' 'Louisville' 'Lawrence' 'Canton'
'New Rochelle' 'Gastonia' 'Jacksonville' 'Auburn' 'Norman' 'Park Ridge'
'Amarillo' 'Lindenhurst' 'Huntsville' 'Fayetteville' 'Costa Mesa'
'Parker' 'Atlanta' 'Gladstone' 'Great Falls' 'Lakeland' 'Montgomery'
'Mesa' 'Green Bay' 'Anaheim' 'Marysville' 'Salem' 'Laredo' 'Grove City'
'Dearborn' 'Warner Robins' 'Vallejo' 'Mission Viejo' 'Rochester Hills'
'Plainfield' 'Sierra Vista' 'Vancouver' 'Cleveland' 'Tyler' 'Burlington'
'Waynesboro' 'Chester' 'Cary' 'Palm Coast' 'Mount Vernon' 'Hialeah'
'Oceanside' 'Evanston' 'Trenton' 'Cottage Grove' 'Bossier City'
'Lancaster' 'Asheville' 'Lake Elsinore' 'Omaha' 'Edmonds' 'Santa Ana'
'Milwaukee' 'Florence' 'Lorain' 'Linden' 'Salinas' 'New Brunswick'
'Garland' 'Norwich' 'Alexandria' 'Toledo' 'Farmington' 'Riverside'
'Torrance' 'Round Rock' 'Boca Raton' 'Virginia Beach' 'Murrieta'
'Olympia' 'Washington' 'Jefferson City' 'Saint Peters' 'Rockford'
'Brownsville' 'Yonkers' 'Oakland' 'Clinton' 'Encinitas' 'Roswell'
'Jonesboro' 'Antioch' 'Homestead' 'La Porte' 'Lansing' 'Cuyahoga Falls'
'Reno' 'Harrisonburg' 'Escondido' 'Royal Oak' 'Rockville' 'Coral Springs'
'Buffalo' 'Boynton Beach' 'Gulfport' 'Fresno' 'Greenville' 'Macon'
'Cedar Rapids' 'Providence' 'Pueblo' 'Deltona' 'Murray' 'Middletown'
'Freeport' 'Pico Rivera' 'Provo' 'Pleasant Grove' 'Smyrna' 'Parma'
'Mobile' 'New Bedford' 'Irving' 'Vineland' 'Glendale' 'Niagara Falls'
'Thomasville' 'Westminster' 'Coppell' 'Pomona' 'North Las Vegas'
'Allentown' 'Tempe' 'Laguna Niguel' 'Bridgeton' 'Everett' 'Watertown'
'Appleton' 'Bellevue' 'Allen' 'El Paso' 'Grapevine' 'Carrollton' 'Kent'
'Lafayette' 'Tigard' 'Skokie' 'Plano' 'Suffolk' 'Indianapolis' 'Bayonne'
'Greensboro' 'Baltimore' 'Kenosha' 'Olathe' 'Tulsa' 'Redmond' 'Raleigh'
'Muskogee' 'Meriden' 'Bowling Green' 'South Bend' 'Spokane' 'Keller'
```

```
'Port Orange' 'Medford' 'Charlottesville' 'Missoula' 'Apopka' 'Reading'
'Broomfield' 'Paterson' 'Oklahoma City' 'Chesapeake' 'Lubbock'
'Johnson City' 'San Bernardino' 'Leominster' 'Bozeman' 'Perth Amboy'
'Ontario' 'Rancho Cucamonga' 'Moorhead' 'Mesquite' 'Stockton'
'Ormond Beach' 'Sunnyvale' 'York' 'College Station' 'Saint Louis'
'Manteca' 'San Angelo' 'Salt Lake City' 'Knoxville' 'Little Rock'
'Lincoln Park' 'Marion' 'Littleton' 'Bangor' 'Southaven' 'New Castle'
'Midland' 'Sioux Falls' 'Fort Collins' 'Clarksville' 'Sacramento'
'Thousand Oaks' 'Malden' 'Holyoke' 'Albuquerque' 'Sparks' 'Coachella'
'Fitchburg' 'Barnstable' 'North Charleston' 'Mount Vernon' 'Tomball'
```

df.describe()

	Sales	Quantity	Discount	Profit
count	9977.000000	9977.000000	9977.000000	9977.00000
mean	230.148902	3.790719	0.156278	28.69013
std	623.721409	2.226657	0.206455	234.45784
min	0.444000	1.000000	0.000000	-6599.97800
25%	17.300000	2.000000	0.000000	1.72620
50%	54.816000	3.000000	0.200000	8.67100
75%	209.970000	5.000000	0.200000	29.37200
max	22638.480000	14.000000	0.800000	8399.97600

df.info()

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 9977 entries, 0 to 9993
Data columns (total 12 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Ship Mode    9977 non-null   object
1   Segment      9977 non-null   object
2   Country       9977 non-null   object
3   City          9977 non-null   object
4   State         9977 non-null   object
5   Region        9977 non-null   object
6   Category      9977 non-null   object
7   Sub-Category  9977 non-null   object
8   Sales         9977 non-null   float64
9   Quantity      9977 non-null   int64
10  Discount      9977 non-null   float64
11  Profit        9977 non-null   float64
dtypes: float64(3), int64(1), object(8)
memory usage: 1013.3+ KB
```

HANDLING MISSING VALUES

df.dropna()

	Ship Mode	Segment	Country	City	State	Region	Category	Sub-Category	Sales	Quantity	Discount	Profit
0	Second Class	Consumer	United States	Henderson	Kentucky	South	Furniture	Bookcases	261.9600	2	0.00	41.9131
1	Second Class	Consumer	United States	Henderson	Kentucky	South	Furniture	Chairs	731.9400	3	0.00	219.5821
2	Second Class	Corporate	United States	Los Angeles	California	West	Office Supplies	Labels	14.6200	2	0.00	6.8714
3	Standard Class	Consumer	United States	Fort Lauderdale	Florida	South	Furniture	Tables	957.5775	5	0.45	-383.0311
4	Standard Class	Consumer	United States	Fort Lauderdale	Florida	South	Office Supplies	Storage	22.3680	2	0.20	2.5164
...
9989	Second Class	Consumer	United States	Miami	Florida	South	Furniture	Furnishings	25.2480	3	0.20	4.1021
9990	Standard Class	Consumer	United States	Costa Mesa	California	West	Furniture	Furnishings	91.9600	2	0.00	15.6331

df.drop_duplicates(inplace=True)

```
df.isna().sum()
```

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

STATISTICAL ANALYSIS

```
df.groupby("Region")["Sales"].sum()
```

```
Region
Central    500724.2708
East       677843.9080
South      391721.9050
West       724309.2935
Name: Sales, dtype: float64
```

```
df.groupby("Region")["Sales"].mean()
```

```
Region
Central    216.295581
East       239.351662
South      241.803645
West       227.985299
Name: Sales, dtype: float64
```

```
df.groupby("Region")["Profit"].sum()
```

```
Region
Central    39662.0474
East       91471.0909
South      46749.4303
West       108214.9893
Name: Profit, dtype: float64
```

```
df.groupby("Region")["Profit"].mean()
```

```
Region
Central    17.132634
East       32.299114
South      28.857673
West       34.062005
Name: Profit, dtype: float64
```

```
df.groupby("Segment")["Sales"].sum()
```

```
Segment
Consumer    1.159418e+06
Corporate    7.059702e+05
Home Office  4.292117e+05
Name: Sales, dtype: float64
```

```
df.groupby("Segment")["Sales"].mean()
```

```
Segment
Consumer    224.693317
Corporate    234.697538
Home Office  241.673237
Name: Sales, dtype: float64
```

```
df.groupby("Segment")["Profit"].sum()
```

```
Segment
Consumer    133939.5218
Corporate    91911.1606
Home Office  60246.8755
Name: Profit, dtype: float64
```

```
df.groupby("Segment")["Profit"].mean()
```

```
Segment
Consumer      25.957272
Corporate      30.555572
Home Office    33.922790
Name: Profit, dtype: float64
```

```
df.groupby("Category")["Sales"].sum()
```

```
Category
Furniture      740568.6663
Office Supplies 718256.0700
Technology      835774.6410
Name: Sales, dtype: float64
```

```
df.groupby("Category")["Sales"].mean()
```

```
Category
Furniture      350.316304
Office Supplies 119.989320
Technology      453.240044
Name: Sales, dtype: float64
```

```
df.groupby("Sub-Category")["Sales"].sum()
```

```
Sub-Category
Accessories    167303.3020
Appliances     107532.1610
Art            27065.7320
Binders        203328.8590
Bookcases      114166.9293
Chairs         327777.7610
Copiers        149528.0300
Envelopes      16476.4020
Fasteners       3024.2800
Furnishings    91658.4440
Labels         12385.7640
Machines       189238.6310
Paper          77960.8940
Phones         329704.6780
Storage        223808.4400
Supplies       46673.5380
Tables         206965.5320
Name: Sales, dtype: float64
```

```
df.groupby("Sub-Category")["Profit"].sum()
```

```
Sub-Category
Accessories    41920.9052
Appliances     18138.0054
Art            6511.8388
Binders        30205.2603
Bookcases     -3398.7542
Chairs        26567.1278
Copiers       55617.8249
Envelopes     6964.1767
Fasteners      949.5182
Furnishings   13042.0534
Labels        5499.1558
Machines      3384.7569
Paper        33830.0381
Phones       44493.0524
Storage      21287.1788
Supplies     -1189.0995
Tables      -17725.4811
Name: Profit, dtype: float64
```

```
df.groupby("Sub-Category")["Sales"].mean()
```

```
Sub-Category
Accessories    216.433767
Appliances     230.755710
Art            34.130810
Binders        133.945230
Bookcases      505.163404
Chairs         532.971969
Copiers        2198.941618
Envelopes       64.867724
Fasteners       13.936774
Furnishings     96.078034
Labels          34.500735
Machines       1645.553313
Paper           58.006618
Phones         371.289052
Storage        264.862059
Supplies       245.650200
```

```
Tables          648.794771
Name: Sales, dtype: float64
```

```
df.groupby("Sub-Category")["Profit"].mean()
```

```
Sub-Category
Accessories    54.231443
Appliances     38.922758
Art            8.211650
Binders       19.898063
Bookcases    -15.038735
Chairs       43.198582
Copiers     817.909190
Envelopes    27.418019
Fasteners     4.375660
Furnishings  13.670916
Labels       15.317983
Machines     29.432669
Paper       25.171159
Phones      50.104789
Storage     25.191928
Supplies    -6.258418
Tables     -55.565771
Name: Profit, dtype: float64
```

```
df.groupby("State")["Sales"].sum()
```

```
State
Alabama      19510.6400
Arizona      35282.0010
Arkansas     11678.1300
California   456629.9285
Colorado     32108.1180
Connecticut  13384.3570
Delaware     27451.0690
District of Columbia  2865.0200
Florida     89473.7080
Georgia     49095.8400
Idaho        4382.4860
Illinois     80127.3690
Indiana     53555.3600
Iowa         4579.7600
Kansas       2914.3100
Kentucky    36591.7500
Louisiana    9217.0300
Maine        1270.5300
Maryland    23705.5230
Massachusetts  28634.4340
Michigan     75879.6440
Minnesota    29863.1500
Mississippi  10771.3400
Missouri     22205.1500
Montana      5589.3520
Nebraska     7464.9300
Nevada       16729.1020
New Hampshire  7292.5240
New Jersey   35764.3120
New Mexico   4783.5220
New York     310349.2150
North Carolina  55603.1640
North Dakota   919.9100
Ohio         77976.7640
Oklahoma     19683.3900
Oregon       17420.7820
Pennsylvania 116383.0100
Rhode Island  22627.9560
South Carolina  8481.7100
South Dakota  1315.5600
Tennessee    30661.8730
Texas        170101.1278
Utah         11220.0560
Vermont       8929.3700
Virginia     70636.7200
Washington   138560.8100
West Virginia 1209.8240
Wisconsin     32114.6100
Wyoming      1603.1360
Name: Sales, dtype: float64
```

```
df.groupby("State")["Sales"].mean()
```

```
State
Alabama      319.846557
Arizona      157.508933
Arkansas     194.635500
California   230.621176
Colorado     176.418231
```

Connecticut	163.223866
Delaware	285.948635
District of Columbia	286.502000
Florida	233.612815
Georgia	266.825217
Idaho	208.689810
Illinois	163.525243
Indiana	359.431946
Iowa	152.658667
Kansas	121.429583
Kentucky	263.250000
Louisiana	219.453095
Maine	158.816250
Maryland	225.766886
Massachusetts	212.106919
Michigan	298.738756
Minnesota	335.541011
Mississippi	203.232830
Missouri	336.441667
Montana	372.623467
Nebraska	196.445526
Nevada	428.951333
New Hampshire	270.093481
New Jersey	275.110092
New Mexico	129.284378
New York	277.345143
North Carolina	223.305880
North Dakota	131.415714
Ohio	166.617017
Oklahoma	298.233182
Oregon	141.632374
Pennsylvania	200.314991
Rhode Island	404.070643
South Carolina	201.945476
South Dakota	109.630000
Tennessee	167.551219
Texas	173.572579
Utah	211.699170
Vermont	811.760909
Virginia	315.342500
Washington	276.017550
West Virginia	302.456000
Wisconsin	291.951000
Wyoming	1603.136000

Name: Sales, dtype: float64

```
df.groupby("State")["Profit"].sum()
```

State	
Alabama	5786.8253
Arizona	-3427.9246
Arkansas	4008.6871
California	76215.9705
Colorado	-6527.8579
Connecticut	3511.4918
Delaware	9977.3748
District of Columbia	1059.5893
Florida	-3399.3017
Georgia	16250.0433
Idaho	826.7231
Illinois	-12593.2976
Indiana	18382.9363
Iowa	1183.8119
Kansas	836.4435
Kentucky	11199.6966
Louisiana	2196.1023
Maine	454.4862
Maryland	7031.1788
Massachusetts	6785.5016
Michigan	24428.0903
Minnesota	10823.1874
Mississippi	3172.9762
Missouri	6436.2105
Montana	1833.3285
Nebraska	2037.0942
Nevada	3316.7659
New Hampshire	1706.5028
New Jersey	9772.9138
New Mexico	1157.1161
New York	74006.1552
North Carolina	-7490.9122
North Dakota	230.1497
Ohio	-16959.3178
Oklahoma	4853.9560
Oregon	-1194.0993
Pennsylvania	-15591.3148
Rhode Island	7285.6293
South Carolina	1769.0566
South Dakota	394.8283


```

Tennessee      -5341.6936
Texas           -25753.1635
Utah            2546.5335
Vermont         2244.9783
Virginia        18597.9504
Washington      33368.2375
West Virginia   185.9216
Wisconsin       8401.8004
Wyoming         100.1960
Name: Profit, dtype: float64

```

```
df.groupby("State")["Profit"].mean()
```

```

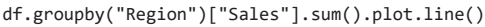
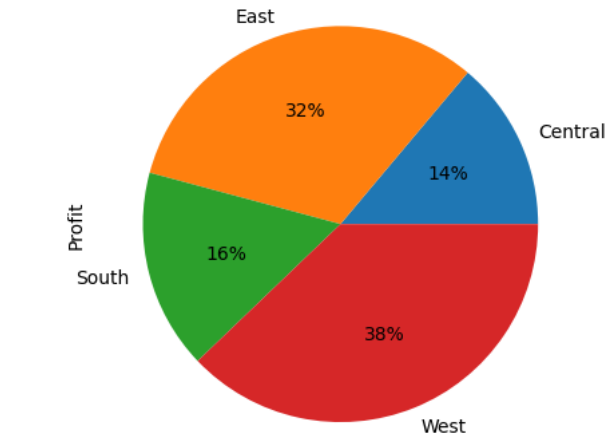
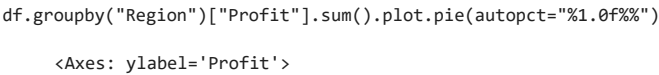
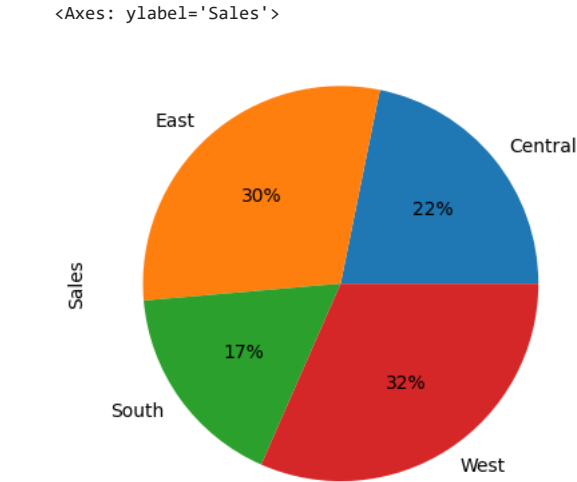
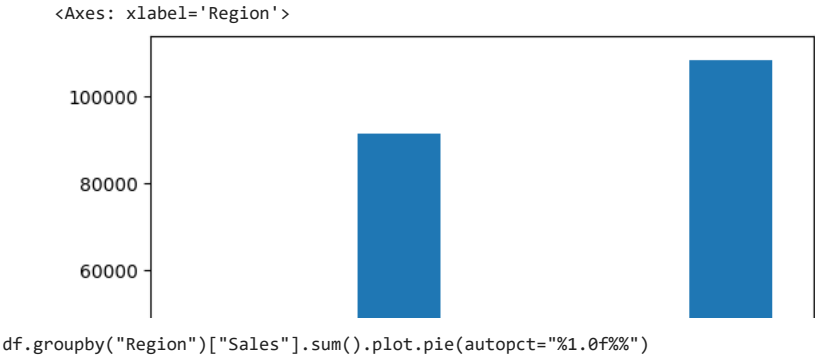
State
Alabama      94.865989
Arizona      -15.303235
Arkansas      66.811452
California    38.492914
Colorado     -35.867351
Connecticut   42.823071
Delaware     103.930988
District of Columbia  105.958930
Florida      -8.875461
Georgia       88.315453
Idaho         39.367767
Illinois     -25.700607
Indiana      123.375411
Iowa         39.460397
Kansas       34.851813
Kentucky     80.573357
Louisiana    52.288150
Maine        56.810775
Maryland     66.963608
Massachusetts  50.262975
Michigan     96.173584
Minnesota    121.608847
Mississippi   59.867475
Missouri     97.518341
Montana     122.221900
Nebraska     53.607742
Nevada       85.045279
New Hampshire  63.203807
New Jersey   75.176260
New Mexico   31.273408
New York     66.135974
North Carolina -30.083985
North Dakota  32.878529
Ohio        -36.237859
Oklahoma     73.544788
Oregon       -9.708124
Pennsylvania -26.835309
Rhode Island 130.100523
South Carolina  42.120395
South Dakota  32.902358
Tennessee   -29.189583
Texas       -26.278738
Utah         48.047802
Vermont     204.088936
Virginia     83.026564
Washington   66.470593
West Virginia  46.480400
Wisconsin    76.380004
Wyoming     100.196000
Name: Profit, dtype: float64

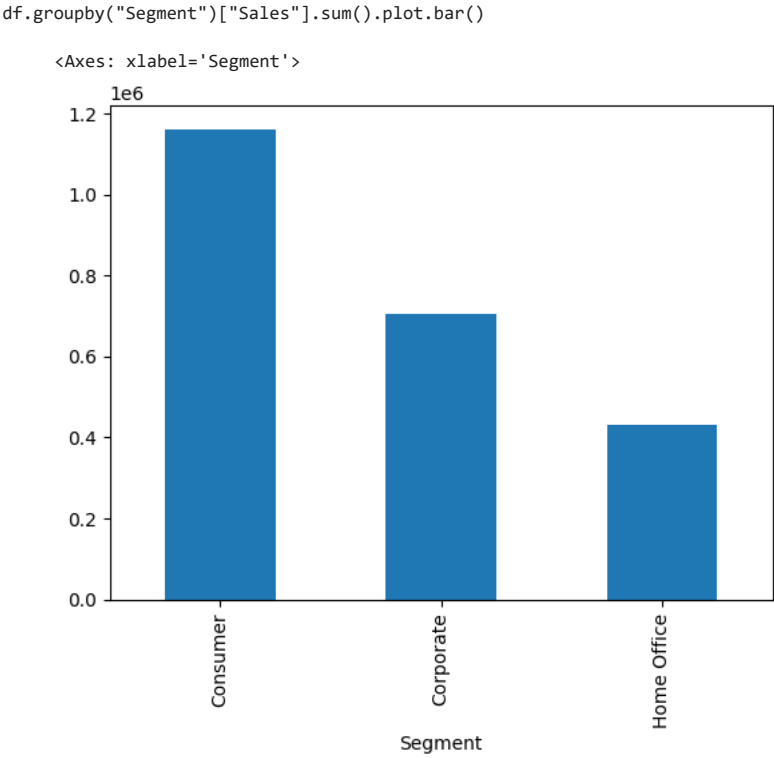
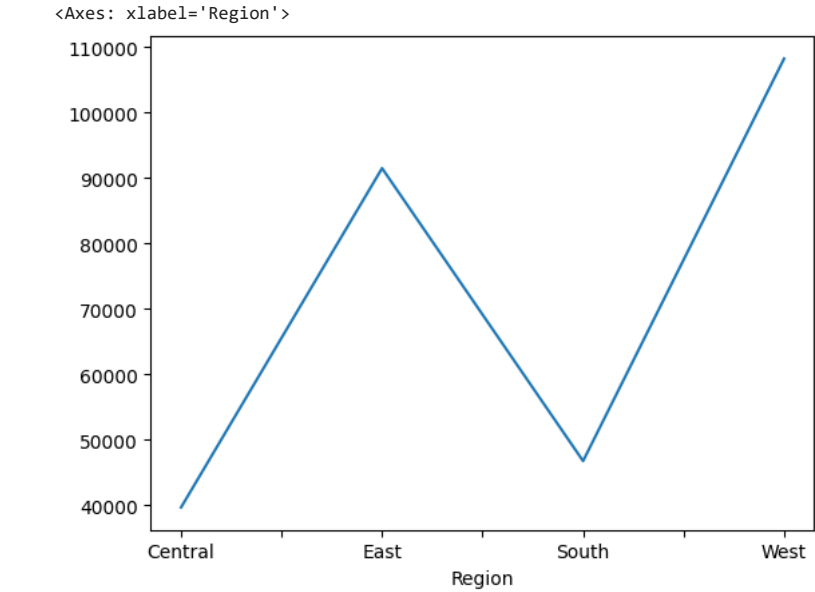
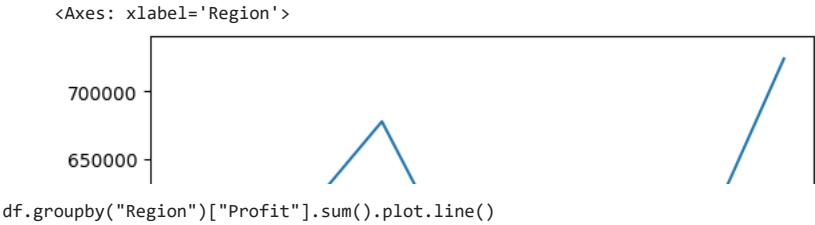
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VISUALIZATIONS

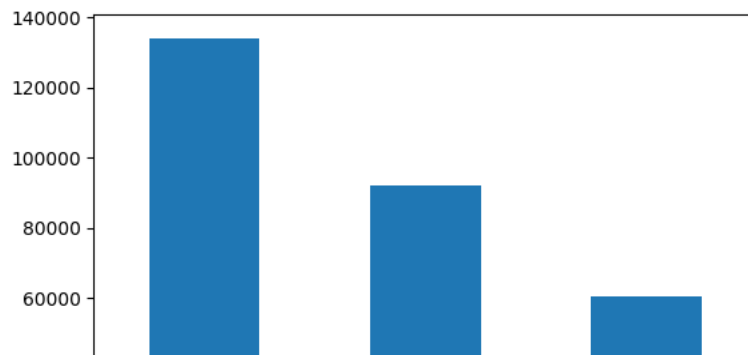
```
df.groupby("Region")["Sales"].sum().plot.bar()
```

```
df.groupby("Region")["Profit"].sum().plot.bar()
```



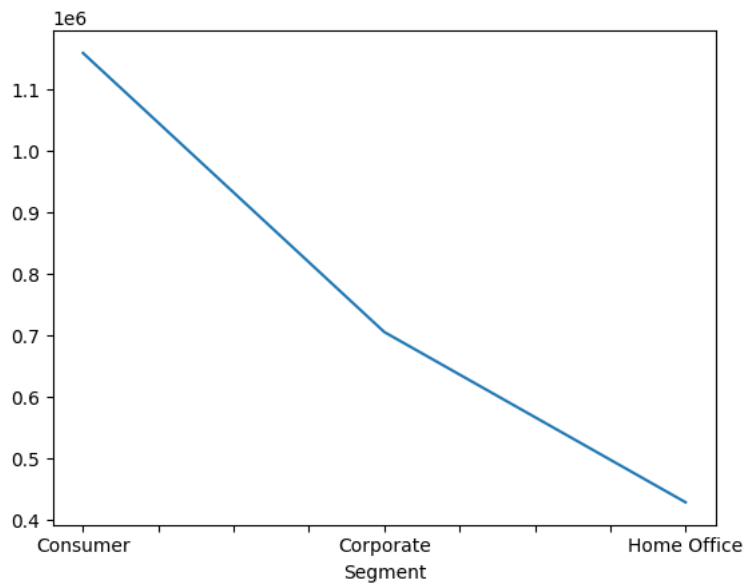


<Axes: xlabel='Segment'>



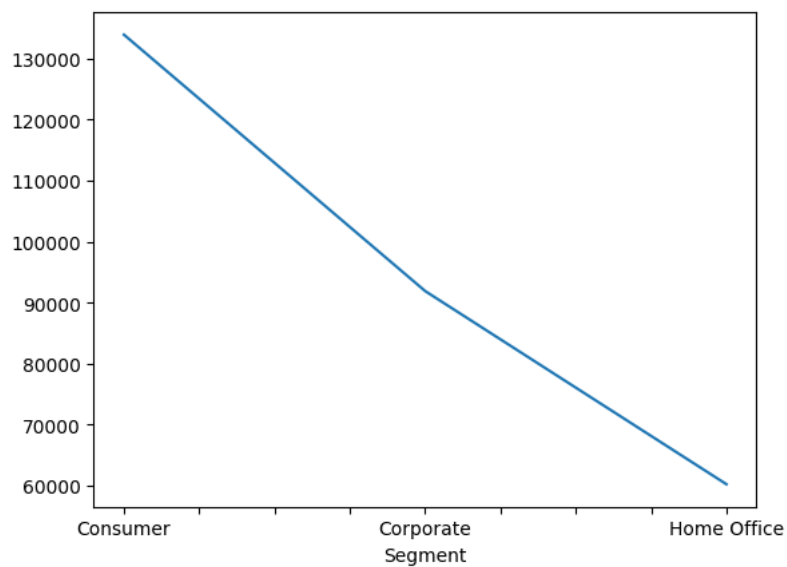
```
df.groupby("Segment")["Sales"].sum().plot.line()
```

<Axes: xlabel='Segment'>



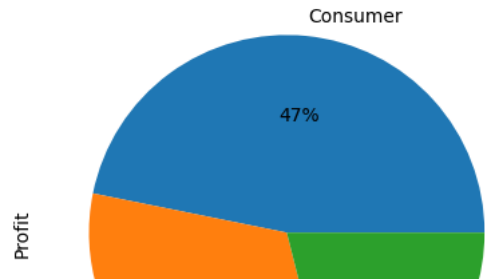
```
df.groupby("Segment")["Profit"].sum().plot.line()
```

<Axes: xlabel='Segment'>



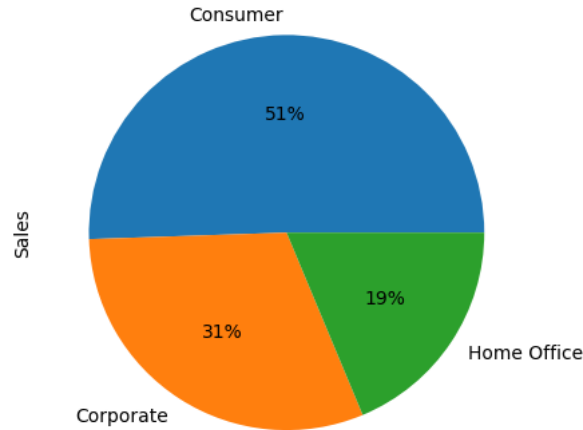
```
df.groupby("Segment")["Profit"].sum().plot.pie(autopct="%1.0f%%")
```

<Axes: ylabel='Profit'>



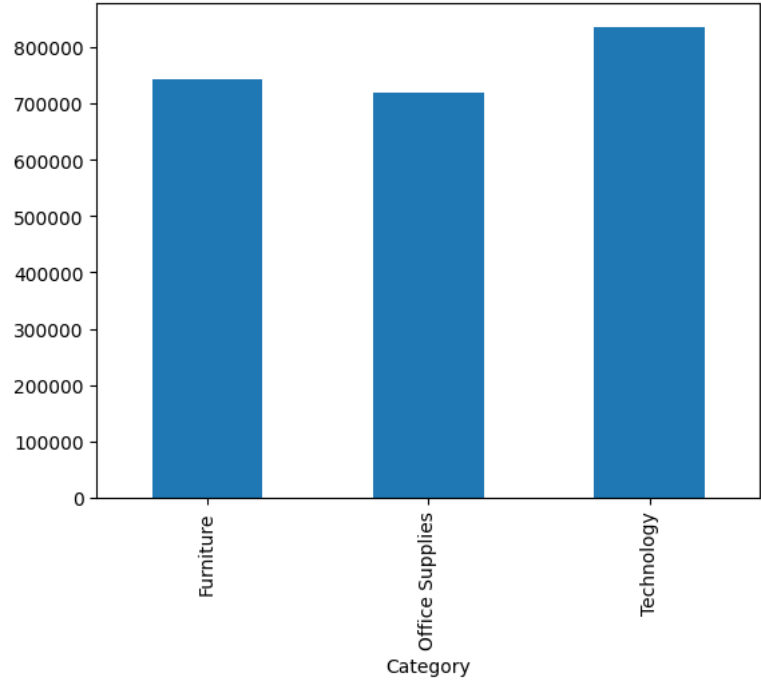
```
df.groupby("Segment")["Sales"].sum().plot.pie(autopct="%1.0f%%")
```

<Axes: ylabel='Sales'>

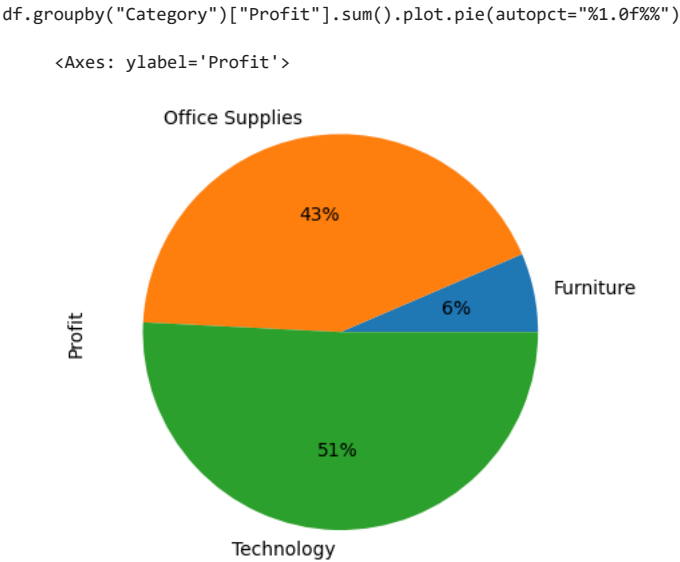
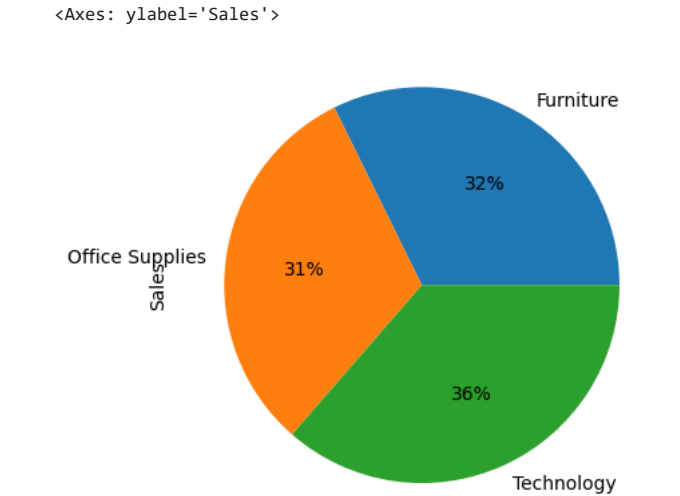
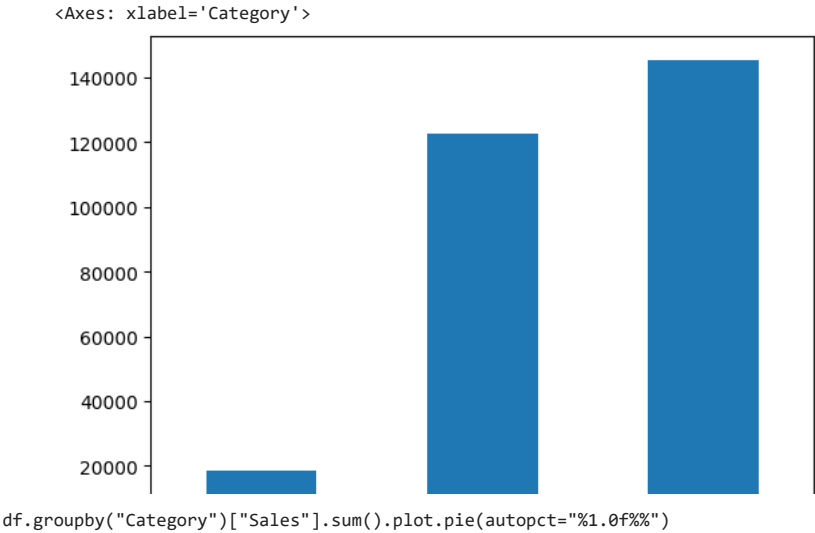


```
df.groupby("Category")["Sales"].sum().plot.bar()
```

<Axes: xlabel='Category'>



```
df.groupby("Category")["Profit"].sum().plot.bar()
```



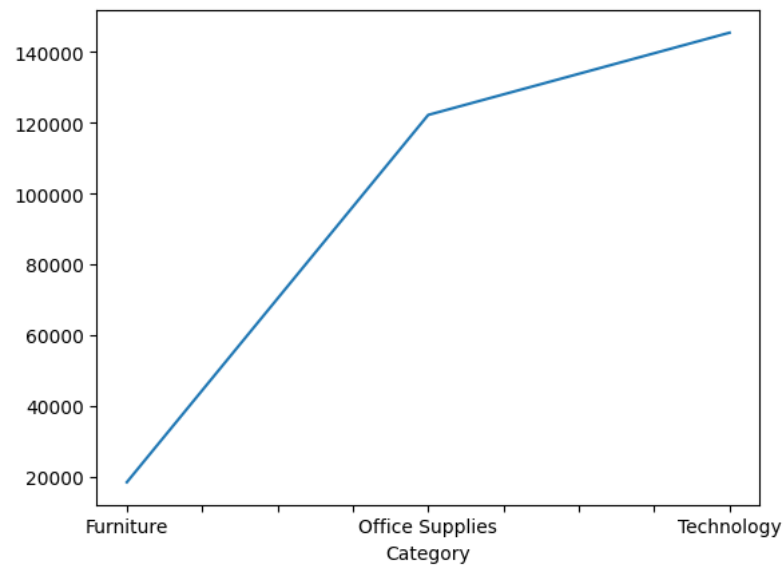
df.groupby("Category")["Sales"].sum().plot.line()

```
<Axes: xlabel='Category'>
```



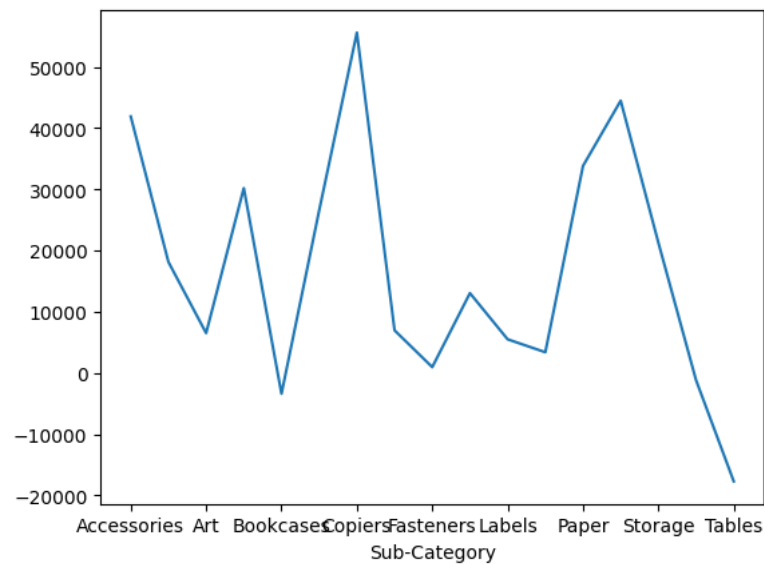
```
df.groupby("Category")["Profit"].sum().plot.line()
```

```
<Axes: xlabel='Category'>
```



```
df.groupby("Sub-Category")["Profit"].sum().plot.line()
```

```
<Axes: xlabel='Sub-Category'>
```



```
df.groupby("Sub-Category")["Sales"].sum().plot.line()
```

