

HNG

March 3, 2025

```
[140]: import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
import seaborn as sns
```

```
[3]: data_df = pd.read_csv(r"C:\Users\essie\Desktop\HNG\SampleSuperstore.csv")
```

```
[4]: data_df.head()
```

```
[4]:
```

	Ship Mode	Segment	Country	City	State	\
0	Second Class	Consumer	United States	Henderson	Kentucky	
1	Second Class	Consumer	United States	Henderson	Kentucky	
2	Second Class	Corporate	United States	Los Angeles	California	
3	Standard Class	Consumer	United States	Fort Lauderdale	Florida	
4	Standard Class	Consumer	United States	Fort Lauderdale	Florida	

	Postal Code	Region	Category	Sub-Category	Sales	Quantity	\
0	42420	South	Furniture	Bookcases	261.9600	2	
1	42420	South	Furniture	Chairs	731.9400	3	
2	90036	West	Office Supplies	Labels	14.6200	2	
3	33311	South	Furniture	Tables	957.5775	5	
4	33311	South	Office Supplies	Storage	22.3680	2	

	Discount	Profit
0	0.00	41.9136
1	0.00	219.5820
2	0.00	6.8714
3	0.45	-383.0310
4	0.20	2.5164

```
[5]: data_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

```

```
[7]: data_df['Postal Code'] = data_df['Postal Code'].astype(object)
```

```
[8]: data_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  object
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(1), object(9)
memory usage: 1015.1+ KB

```

```
[9]: data_df.describe()
```

```

[9]:
count      Sales      Quantity      Discount      Profit
count  9994.000000  9994.000000  9994.000000  9994.000000
mean    229.858001    3.789574    0.156203    28.656896
std     623.245101    2.225110    0.206452   234.260108
min       0.444000    1.000000    0.000000  -6599.978000
25%     17.280000    2.000000    0.000000    1.728750

```

50%	54.490000	3.000000	0.200000	8.666500
75%	209.940000	5.000000	0.200000	29.364000
max	22638.480000	14.000000	0.800000	8399.976000

```
[11]: if data_df.duplicated().sum() > 0:
      print('duplicates found')
      else:
      print('no duplicates')
```

duplicates found

```
[126]: # 1. Check for Duplicates
duplicates = data_df.duplicated()
print(f"Number of duplicate rows: {duplicates.sum()}")

# Optional: Display duplicate rows
if duplicates.sum() > 0:
    print("Duplicate rows:")
    print(data_df[duplicates])
```

Number of duplicate rows: 17

Duplicate rows:

	Ship Mode	Segment	Country	City	State \
950	Standard Class	Home Office	United States	Philadelphia	Pennsylvania
3406	Standard Class	Home Office	United States	Columbus	Ohio
3670	Standard Class	Consumer	United States	Salem	Oregon
4117	Standard Class	Consumer	United States	Los Angeles	California
4553	Standard Class	Consumer	United States	San Francisco	California
5905	Same Day	Home Office	United States	San Francisco	California
6146	Standard Class	Corporate	United States	San Francisco	California
6334	Standard Class	Consumer	United States	New York City	New York
6357	Standard Class	Corporate	United States	Seattle	Washington
7608	Standard Class	Consumer	United States	San Francisco	California
7735	Standard Class	Corporate	United States	Seattle	Washington
7759	Standard Class	Corporate	United States	Houston	Texas
8032	First Class	Consumer	United States	Houston	Texas
8095	Second Class	Consumer	United States	Seattle	Washington
9262	Standard Class	Consumer	United States	Detroit	Michigan
9363	Standard Class	Home Office	United States	Seattle	Washington
9477	Second Class	Corporate	United States	Chicago	Illinois

	Postal Code	Region	Category	Sub-Category	Sales	Quantity \
950	19120	East	Office Supplies	Paper	15.552	3
3406	43229	East	Furniture	Chairs	281.372	2
3670	97301	West	Office Supplies	Paper	10.368	2
4117	90036	West	Office Supplies	Paper	19.440	3
4553	94122	West	Office Supplies	Paper	12.840	3
5905	94122	West	Office Supplies	Labels	41.400	4

6146	94122	West	Office Supplies	Art	11.760	4
6334	10011	East	Office Supplies	Paper	49.120	4
6357	98103	West	Office Supplies	Paper	25.920	4
7608	94122	West	Office Supplies	Paper	25.920	4
7735	98105	West	Office Supplies	Paper	19.440	3
7759	77041	Central	Office Supplies	Paper	15.552	3
8032	77041	Central	Office Supplies	Paper	47.952	3
8095	98115	West	Office Supplies	Paper	12.960	2
9262	48227	Central	Furniture	Chairs	389.970	3
9363	98105	West	Furniture	Furnishings	22.140	3
9477	60653	Central	Office Supplies	Binders	3.564	3

	Discount	Profit
950	0.2	5.4432
3406	0.3	-12.0588
3670	0.2	3.6288
4117	0.0	9.3312
4553	0.0	5.7780
5905	0.0	19.8720
6146	0.0	3.1752
6334	0.0	23.0864
6357	0.0	12.4416
7608	0.0	12.4416
7735	0.0	9.3312
7759	0.2	5.4432
8032	0.2	16.1838
8095	0.0	6.2208
9262	0.0	35.0973
9363	0.0	6.4206
9477	0.8	-6.2370

```
[127]: cleaned_data = data_df.drop_duplicates()
```

```
[128]: print(f"Number of rows after removing duplicates: {len(cleaned_data)}")
```

Number of rows after removing duplicates: 9977

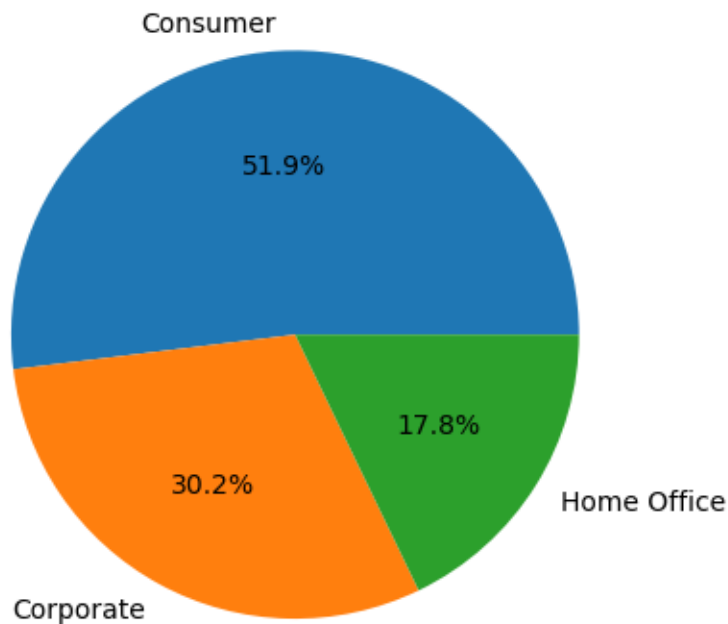
Exploratory Analysis

```
[129]: #number of customers in each segment
no_customers = cleaned_data['Segment'].value_counts().reset_index()
no_customers = no_customers.rename(columns={'Segment':'customer_type', 'count':
↳ 'Total_customers'})
print(no_customers)
```

	customer_type	Total_customers
0	Consumer	5183
1	Corporate	3015
2	Home Office	1779

```
[130]: plt.
        ↳pie(no_customers['Total_customers'],labels=no_customers['customer_type'],autopct='%1.
        ↳1f%%')
plt.title('Distribution of customers by segment')
plt.show()
```

Distribution of customers by segment



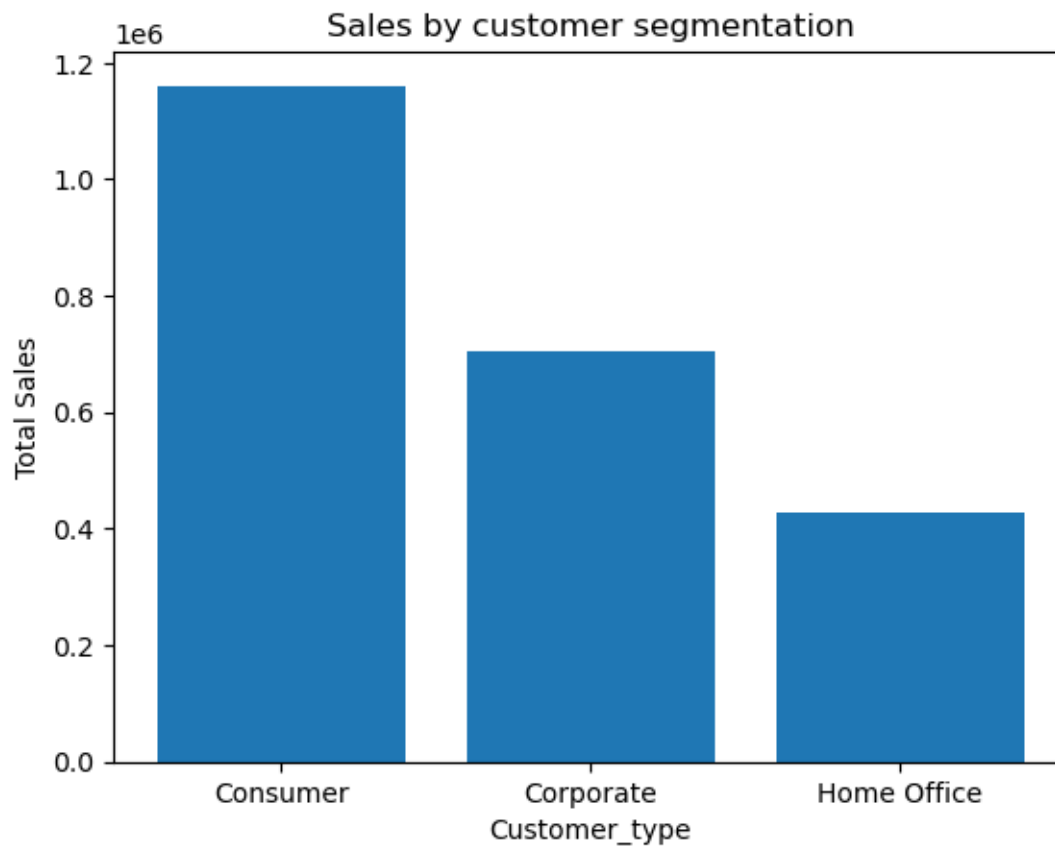
```
[131]: #sales by category
sales_by_category = cleaned_data.groupby('Segment')['Sales'].sum().reset_index()
sales_by_category = sales_by_category.rename(columns={'Segment':
        ↳'Customer_type', 'Sales': 'Total_sales'})
print(sales_by_category)
```

	Customer_type	Total_sales
0	Consumer	1.160833e+06
1	Corporate	7.060701e+05
2	Home Office	4.292927e+05

```
[52]: plt.bar(sales_by_category['Customer_type'], sales_by_category['Total_sales'])
        ↳# x, height

plt.title('Sales by customer segmentation')
plt.xlabel('Customer_type')
```

```
plt.ylabel('Total Sales')
plt.show()
```



shipping Mode

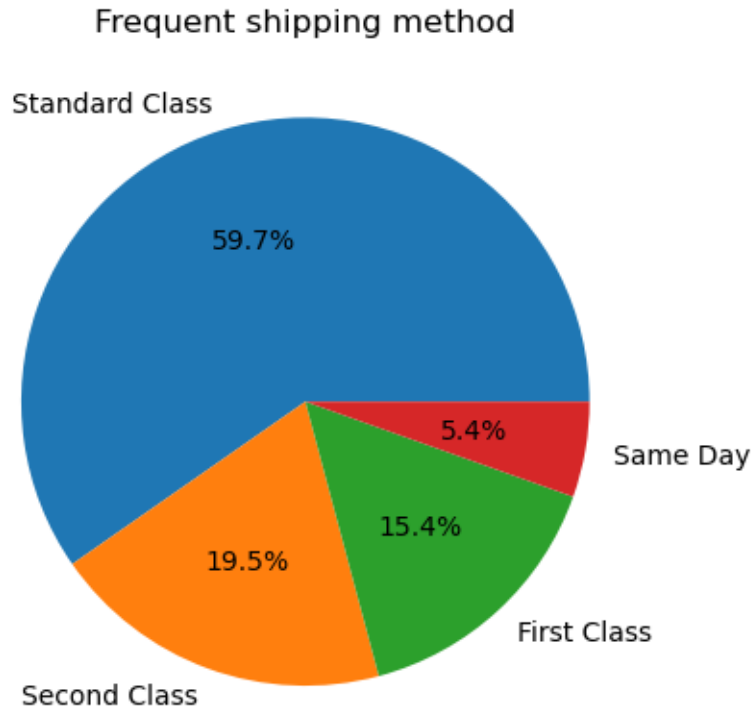
```
[54]: type_of_shipping = cleaned_data['Ship Mode'].unique()
print(type_of_shipping)
```

```
['Second Class' 'Standard Class' 'First Class' 'Same Day']
```

```
[58]: #frequency of use
shipping_mode = cleaned_data['Ship Mode'].value_counts().reset_index()
shipping_mode = shipping_mode.rename(columns={'Ship Mode': 'Mode of shipment', 'count': 'use frequency'})
print(shipping_mode)
```

	Mode of shipment	use frequency
0	Standard Class	5955
1	Second Class	1943
2	First Class	1537
3	Same Day	542

```
[62]: plt.pie(shipping_mode['use frequency'], labels=shipping_mode['Mode of_
↪shipment'], autopct='%1.1f%%')
plt.title('Frequent shipping method')
plt.show()
```



Geogrphical Analysis

```
[132]: States =cleaned_data['State'].value_counts().reset_index()
States = States.rename(columns={'count':'Number of customers'})
print(States.head())
```

	State	Number of customers
0	California	1996
1	New York	1127
2	Texas	983
3	Pennsylvania	586
4	Washington	502

```
[71]: City =cleaned_data['City'].value_counts().reset_index()
City = City.rename(columns={'count':'Number of customers'})
print(City.head())
```

	City	Number of customers
0	New York City	914

1	Los Angeles	746
2	Philadelphia	536
3	San Francisco	506
4	Seattle	424

```
[136]: #Sales by States
State_sales = cleaned_data.groupby(['State'])['Sales'].sum().reset_index()
top_states_sales=State_sales.sort_values(by='Sales',ascending =False)
print(top_states_sales.head(5).reset_index(drop=True))
```

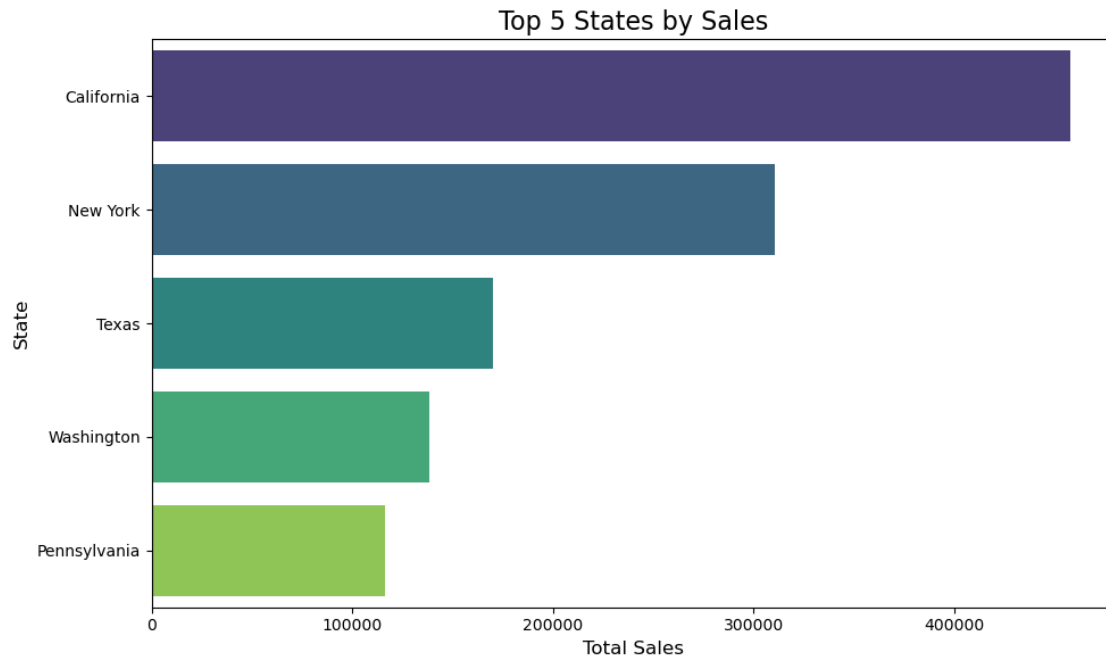
	State	Sales
0	California	457576.2715
1	New York	310827.1510
2	Texas	170124.5418
3	Washington	138560.8100
4	Pennsylvania	116496.3620

```
[141]: import seaborn as sns

top_5_states_sales = top_states_sales.head(5)

plt.figure(figsize=(10, 6))
sns.barplot(
    x='Sales',
    y='State',
    data=top_5_states_sales,
    palette='viridis'
)
plt.title('Top 5 States by Sales', fontsize=16)
plt.xlabel('Total Sales', fontsize=12)
plt.ylabel('State', fontsize=12)

plt.tight_layout()
plt.show()
```

Product analysis

```
[79]: product_category = cleaned_data['Category'].unique()
      print(product_category)
```

```
['Furniture' 'Office Supplies' 'Technology']
```

```
[86]: subcategory_count = cleaned_data.groupby('Category')['Sub-Category'].nunique().
      ↪reset_index()
      subcategory_count = subcategory_count.sort_values(by = 'Sub-Category', ascending=
      ↪False)
      print(subcategory_count.reset_index(drop=True))
```

	Category	Sub-Category	
0	Office Supplies		9
1	Furniture		4
2	Technology		4

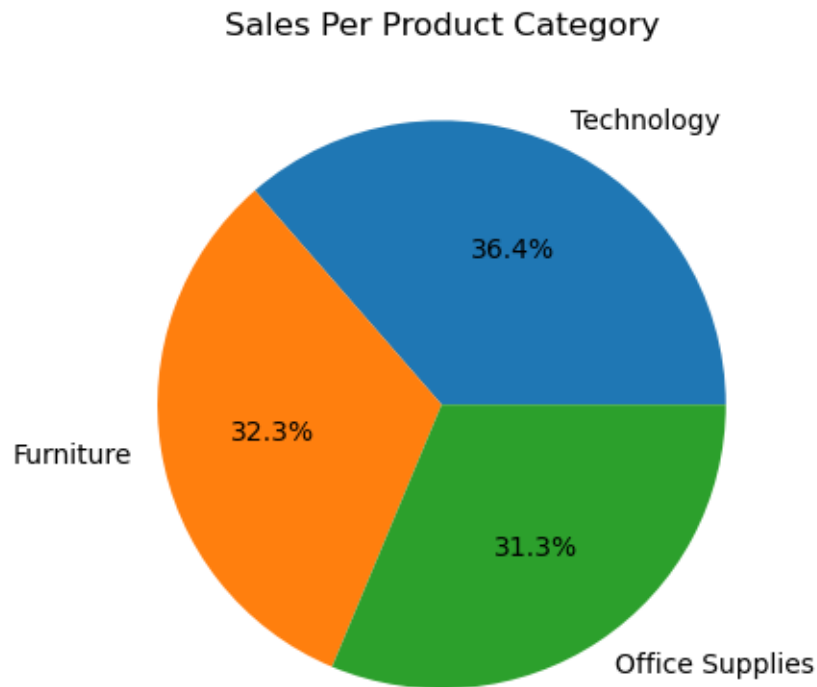
```
[92]: #Sales per category

Sales_per_category = cleaned_data.groupby(['Category'])['Sales'].sum().
      ↪reset_index()
Sales_per_category = Sales_per_category.sort_values(by = 'Sales', ascending =
      ↪False)
print(Sales_per_category.reset_index(drop=True))
```

Category	Sales
----------	-------

```
0      Technology  836154.0330
1      Furniture  741306.3133
2 Office Supplies  718735.2440
```

```
[167]: plt.
        ↳pie(Sales_per_category['Sales'],labels=Sales_per_category['Category'],autopct='%1.
        ↳1f%%')
plt.title('Sales Per Product Category')
plt.show()
```



```
[95]: #Total Sales
Total_sales = cleaned_data['Sales'].sum()
print(Total_sales)
```

```
2296195.5903
```

```
[96]: # Total quantity
Total_quantity =cleaned_data['Quantity'].sum()
print(Total_quantity)
```

```
37820
```

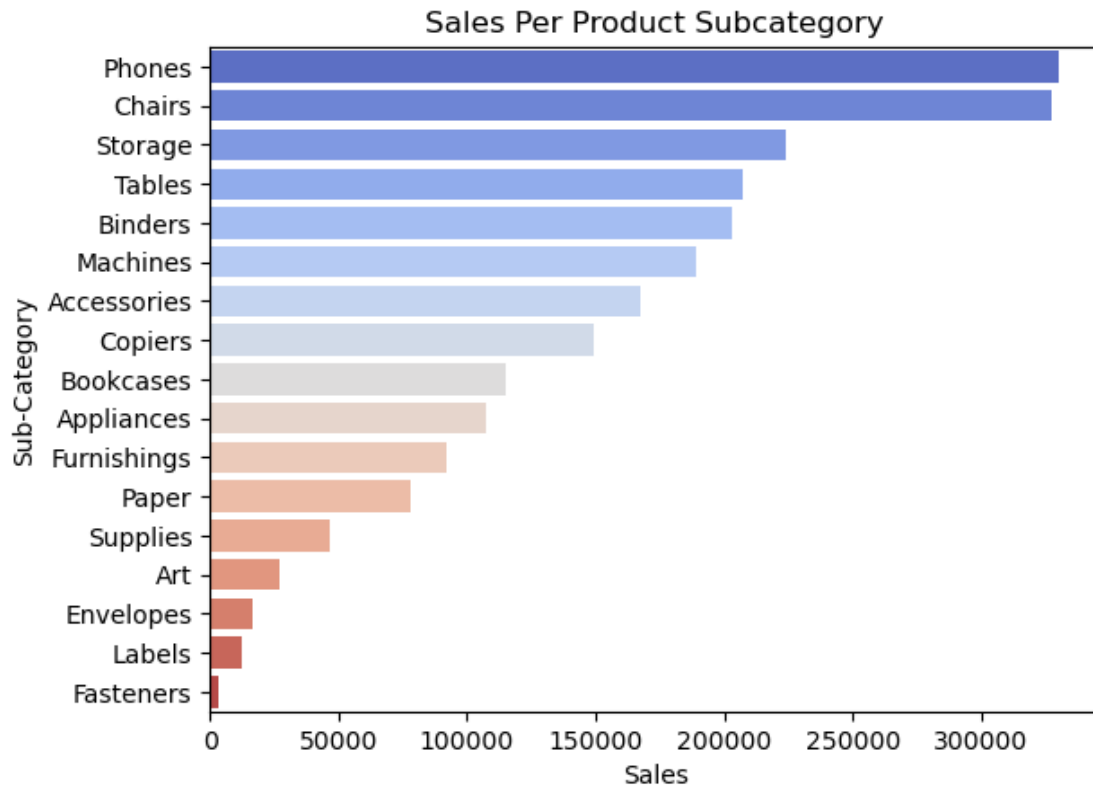
```
[98]: #Total profit
Total_profit = cleaned_data['Profit'].sum()
print(Total_profit)
```

286241.4226

```
[146]: product_subcategory = cleaned_data.groupby(['Sub-Category'])['Sales'].sum().
        ↪reset_index()
Top_product_subcategory = product_subcategory.sort_values(by='Sales', ascending=
        ↪False)
print(Top_product_subcategory.reset_index(drop=True))
```

	Sub-Category	Sales
0	Phones	330007.0540
1	Chairs	327777.7610
2	Storage	223843.6080
3	Tables	206965.5320
4	Binders	203409.1690
5	Machines	189238.6310
6	Accessories	167380.3180
7	Copiers	149528.0300
8	Bookcases	114879.9963
9	Appliances	107532.1610
10	Furnishings	91683.0240
11	Paper	78224.1420
12	Supplies	46673.5380
13	Art	27107.0320
14	Envelopes	16476.4020
15	Labels	12444.9120
16	Fasteners	3024.2800

```
[161]: Top_product_subcategory = product_subcategory.sort_values(by='Sales', ascending=
        ↪False)
plt.
        ↪barh(Top_product_subcategory['Sub-Category'],Top_product_subcategory['Sales'])
plt.xlabel('Sales')
plt.ylabel('Product Subcategory')
plt.title('Sales Per Product Subcategory')
sns.barplot(
    x='Sales',
    y='Sub-Category',
    data=Top_product_subcategory,
    palette='coolwarm'
)
plt.show()
```



```
[162]: product_subcategory = cleaned_data.groupby(['Sub-Category'])['Profit'].sum().
        ↪reset_index()
        Top_product_subcategory = product_subcategory.sort_values(by='Profit',
        ↪ascending =False)
        print(Top_product_subcategory.reset_index(drop=True))
```

	Sub-Category	Profit
0	Copiers	55617.8249
1	Phones	44515.7306
2	Accessories	41936.6357
3	Paper	33944.2395
4	Binders	30228.0003
5	Chairs	26567.1278
6	Storage	21278.8264
7	Appliances	18138.0054
8	Furnishings	13052.7230
9	Envelopes	6964.1767
10	Art	6524.6118
11	Labels	5526.3820
12	Machines	3384.7569
13	Fasteners	949.5182
14	Supplies	-1189.0995

```

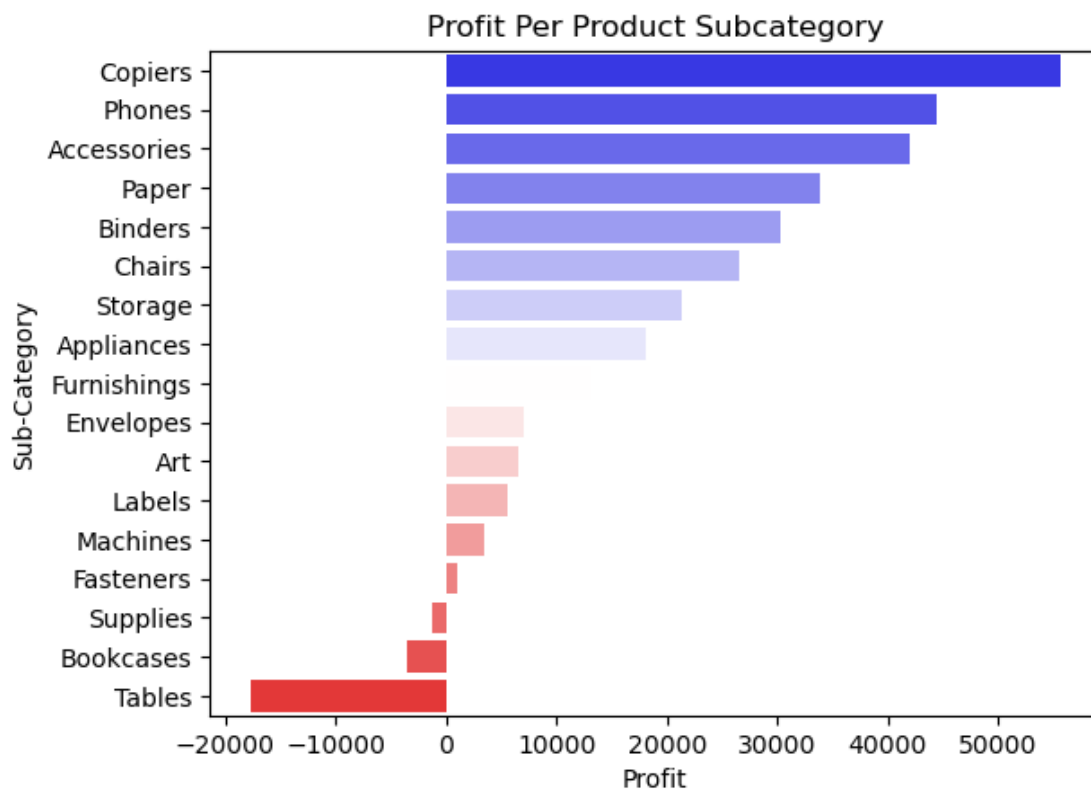
15     Bookcases   -3472.5560
16         Tables -17725.4811

```

```

[164]: Top_product_subcategory = product_subcategory.sort_values(by='Profit',
    ↪ascending =False)
plt.
    ↪barh(Top_product_subcategory['Sub-Category'],Top_product_subcategory['Profit'])
plt.xlabel('Profit')
plt.ylabel('Product Subcategory')
plt.title('Profit Per Product Subcategory')
sns.barplot(
    x='Profit',
    y='Sub-Category',
    data=Top_product_subcategory,
    palette='bwr'
)
plt.show()

```



```

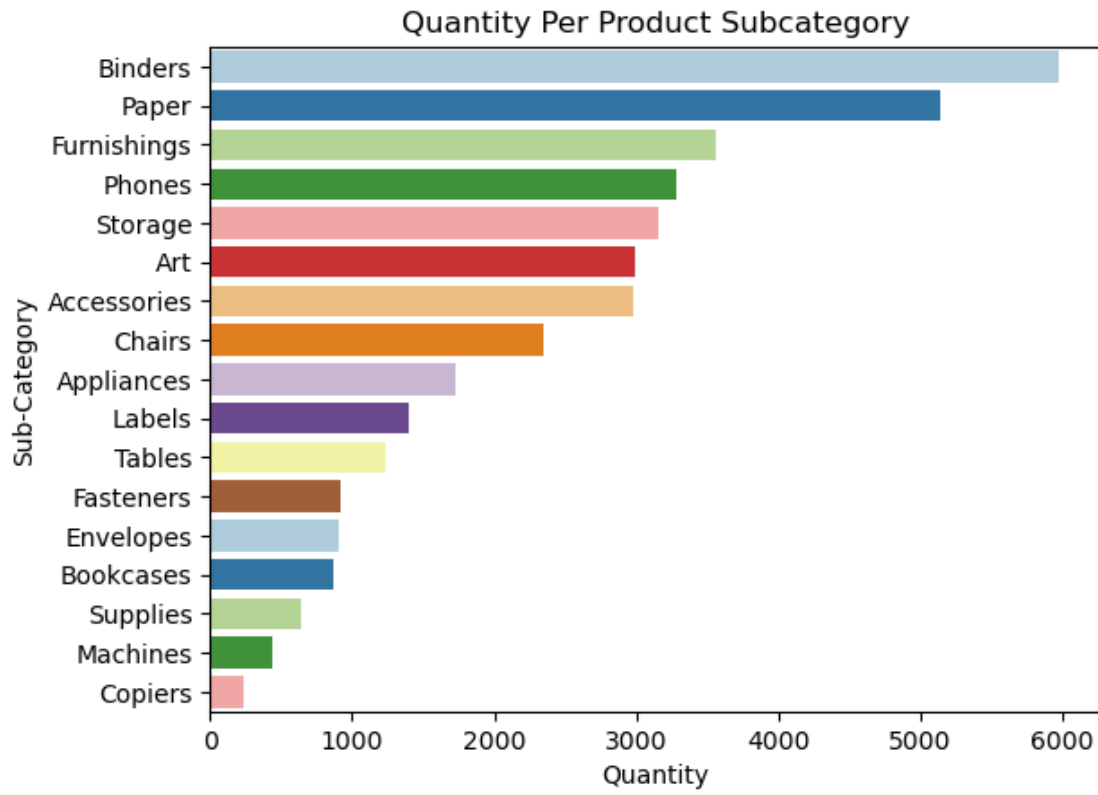
[165]: product_subcategory = cleaned_data.groupby(['Sub-Category'])['Quantity'].sum().
    ↪reset_index()

```

```
Top_product_subcategory = product_subcategory.sort_values(by='Quantity',
↳ascending =False)
print(Top_product_subcategory.reset_index(drop=True))
```

	Sub-Category	Quantity
0	Binders	5971
1	Paper	5144
2	Furnishings	3560
3	Phones	3289
4	Storage	3158
5	Art	2996
6	Accessories	2976
7	Chairs	2351
8	Appliances	1729
9	Labels	1396
10	Tables	1241
11	Fasteners	914
12	Envelopes	906
13	Bookcases	868
14	Supplies	647
15	Machines	440
16	Copiers	234

```
[166]: Top_product_subcategory = product_subcategory.sort_values(by='Quantity',
↳ascending =False)
plt.
↳barh(Top_product_subcategory['Sub-Category'],Top_product_subcategory['Quantity'])
plt.xlabel('Quantity')
plt.ylabel('Product Subcategory')
plt.title('Quantity Per Product Subcategory')
sns.barplot(
    x='Quantity',
    y='Sub-Category',
    data=Top_product_subcategory,
    palette='Paired'
)
plt.show()
```



[]: