

# store-sales-profit-analysis

December 6, 2023

**\*\* Store Sales and Profit Analysis using Python \*\***

```
[ ]: # Now let's start this task by importing necessary libraies:
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
import plotly.graph_objects as go
import plotly.io as pio
import plotly.colors as colors
pio.templates.default = 'plotly_white'
```

```
[ ]: data = pd.read_csv('Sample - Superstore.csv',encoding = 'latin1')
data.head()
```

```
[ ]:
Row ID      Order ID  Order Date  Ship Date      Ship Mode Customer ID \
0         1  CA-2016-152156  11/8/2016  11/11/2016    Second Class  CG-12520
1         2  CA-2016-152156  11/8/2016  11/11/2016    Second Class  CG-12520
2         3  CA-2016-138688  6/12/2016  6/16/2016    Second Class  DV-13045
3         4  US-2015-108966  10/11/2015  10/18/2015  Standard Class  SO-20335
4         5  US-2015-108966  10/11/2015  10/18/2015  Standard Class  SO-20335
```

```
Customer Name  Segment      Country      City ... \
0    Claire Gute  Consumer  United States  Henderson ...
1    Claire Gute  Consumer  United States  Henderson ...
2  Darrin Van Huff  Corporate  United States  Los Angeles ...
3  Sean O'Donnell  Consumer  United States  Fort Lauderdale ...
4  Sean O'Donnell  Consumer  United States  Fort Lauderdale ...
```

```
Postal Code  Region      Product ID      Category Sub-Category \
0    42420    South  FUR-BO-10001798  Furniture  Bookcases
1    42420    South  FUR-CH-10000454  Furniture  Chairs
2    90036    West   OFF-LA-10000240  Office Supplies  Labels
3    33311    South  FUR-TA-10000577  Furniture  Tables
4    33311    South  OFF-ST-10000760  Office Supplies  Storage
```

	Product Name	Sales	Quantity \
0	Bush Somerset Collection Bookcase	261.9600	2
1	Hon Deluxe Fabric Upholstered Stacking Chairs,...	731.9400	3
2	Self-Adhesive Address Labels for Typewriters b...	14.6200	2
3	Bretford CR4500 Series Slim Rectangular Table	957.5775	5
4	Eldon Fold 'N Roll Cart System	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 rows x 21 columns]

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

```
[ ]: Index(['Row ID', 'Order ID', 'Order Date', 'Ship Date', 'Ship Mode',
          'Customer ID', 'Customer Name', 'Segment', 'Country', 'City', 'State',
          'Postal Code', 'Region', 'Product ID', 'Category', 'Sub-Category',
          'Product Name', 'Sales', 'Quantity', 'Discount', 'Profit'],
          dtype='object')
```

```
[ ]: data.isnull().sum()
```

```
[ ]: Row ID      0
      Order ID   0
      Order Date  0
      Ship Date   0
      Ship Mode   0
      Customer ID  0
      Customer Name 0
      Segment     0
      Country     0
      City        0
      State       0
      Postal Code  0
      Region      0
      Product ID  0
      Category    0
      Sub-Category 0
      Product Name 0
      Sales       0
      Quantity    0
      Discount    0
      Profit      0
```

dtype: int64

```
[ ]: # let's start by looking at the descriptive statistics of the dataset
data.describe()
```

```
[ ]:
      Row ID  Postal Code    Sales  Quantity  Discount \
count  9994.000000  9994.000000  9994.000000  9994.000000  9994.000000
mean    4997.500000  55190.379428   229.858001    3.789574    0.156203
std     2885.163629  32063.693350   623.245101    2.225110    0.206452
min         1.000000   1040.000000    0.444000    1.000000    0.000000
25%     2499.250000  23223.000000    17.280000    2.000000    0.000000
50%     4997.500000  56430.500000    54.490000    3.000000    0.200000
75%     7495.750000  90008.000000   209.940000    5.000000    0.200000
max     9994.000000  99301.000000  22638.480000   14.000000    0.800000

      Profit
count  9994.000000
mean    28.656896
std     234.260108
min   -6599.978000
25%      1.728750
50%      8.666500
75%     29.364000
max     8399.976000
```

The dataset has an order date column. we can use this column to create new columns like order month, year and orderday, which will be very valuable for sales and profit analysis according to time periods.

```
[ ]: data['Order Date'].dtypes
```

```
[ ]: dtype('O')
```

```
[ ]: data['Order Date'] = pd.to_datetime(data['Order Date'])
```

```
[ ]: data['Order Date'].dtypes
```

```
[ ]: dtype('<M8[ns]')
```

```
[ ]: data['Ship Date'] = pd.to_datetime(data['Ship Date'])
```

```
[ ]: data['Order Month'] = data['Order Date'].dt.month
```

```
[ ]: data['Order Year'] = data['Order Date'].dt.year
```

```
[ ]: data['Order Day of Week']=data['Order Date'].dt.dayofweek
```

## 1 Now let's have a look at the monthly sales:

```
[ ]: sales_by_month = data.groupby('Order Month')['Sales'].sum().reset_index()
      # Plot the line chart
      plt.plot(sales_by_month['Order Month'], sales_by_month['Sales'])
      plt.xlabel('Order Month')
      plt.ylabel('Sales')
      plt.title('Sales by Month')
      plt.show()
```



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

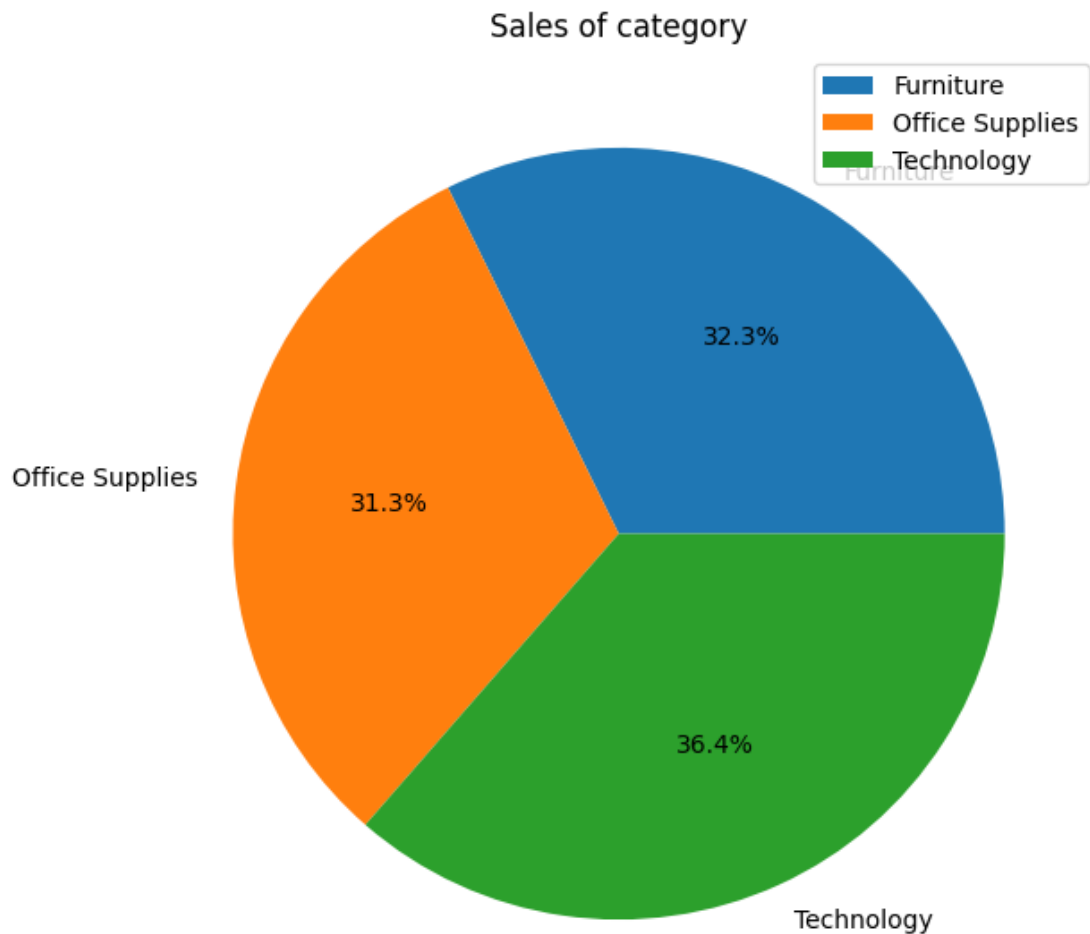
```
[ ]: Office Supplies    6026
      Furniture         2121
      Technology        1847
      Name: Category, dtype: int64
```

```
[ ]: # now let's have a look at the sales by category:
      sales_by_category = data.groupby('Category')['Sales'].sum().reset_index()
      labels = sales_by_category['Category']
```

```

sizes = sales_by_category['Sales']
plt.figure(figsize=(10,7))
plt.pie(sizes, labels = labels,autopct='%1.1f%%')
plt.title('Sales of category')
plt.legend()
plt.show()

```



```
[ ]: data['Sub-Category'].value_counts()
```

```
[ ]: Binders      1523
      Paper       1370
      Furnishings  957
      Phones      889
      Storage     846
      Art         796
```

```

Accessories      775
Chairs           617
Appliances       466
Labels           364
Tables           319
Envelopes        254
Bookcases        228
Fasteners        217
Supplies         190
Machines         115
Copiers          68
Name: Sub-Category, dtype: int64

```

```

[ ]: sales_by_sub_category = data.groupby('Sub-Category')['Sales'].sum().
      ↪reset_index().sort_values(by='Sales',ascending=False)
sales_by_sub_category

```

```

[ ]:
Sub-Category      Sales
13      Phones  330007.0540
5        Chairs  328449.1030
14      Storage  223843.6080
16      Tables  206965.5320
3        Binders  203412.7330
11     Machines  189238.6310
0  Accessories  167380.3180
6        Copiers  149528.0300
4    Bookcases  114879.9963
1    Appliances  107532.1610
9  Furnishings   91705.1640
12         Paper   78479.2060
15     Supplies   46673.5380
2          Art    27118.7920
7    Envelopes   16476.4020
10     Labels    12486.3120
8    Fasteners    3024.2800

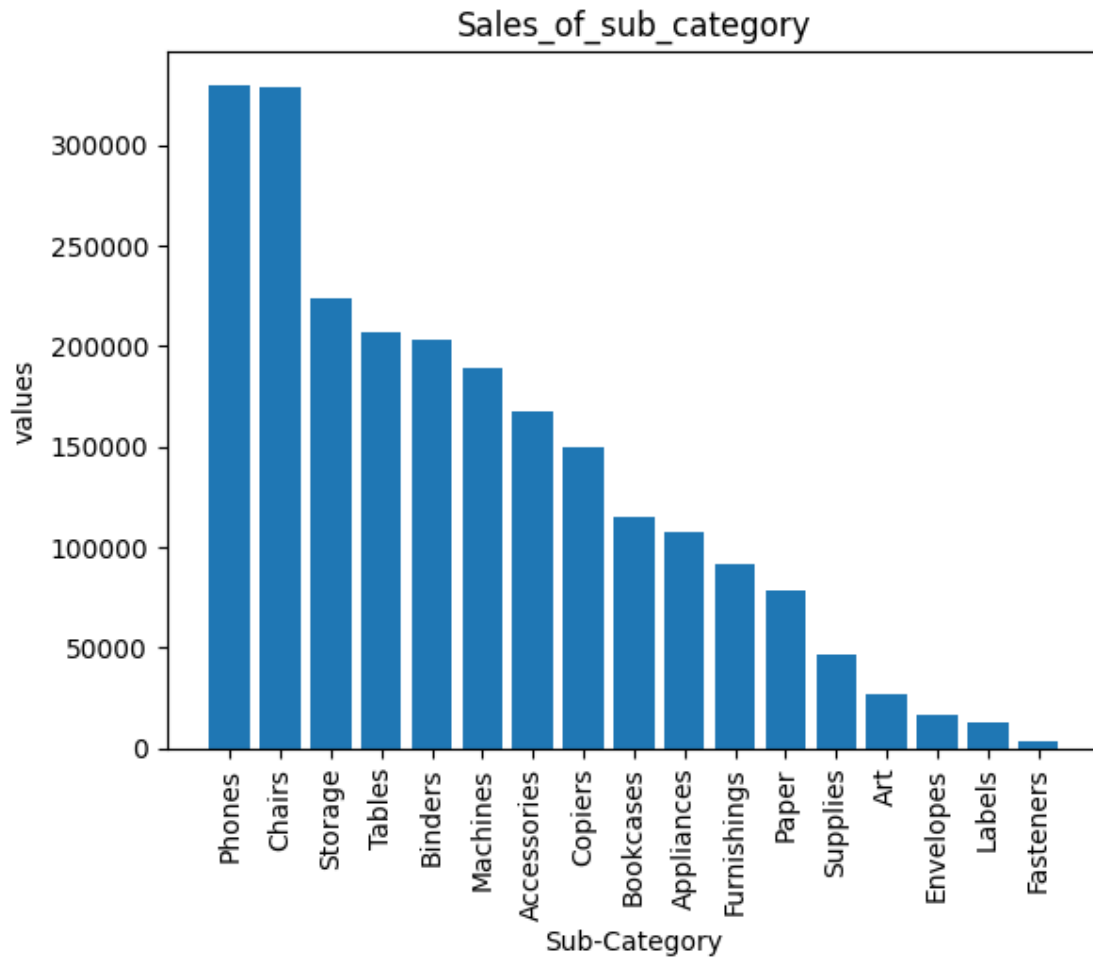
```

Now let's have a look at the sales by sub category

```

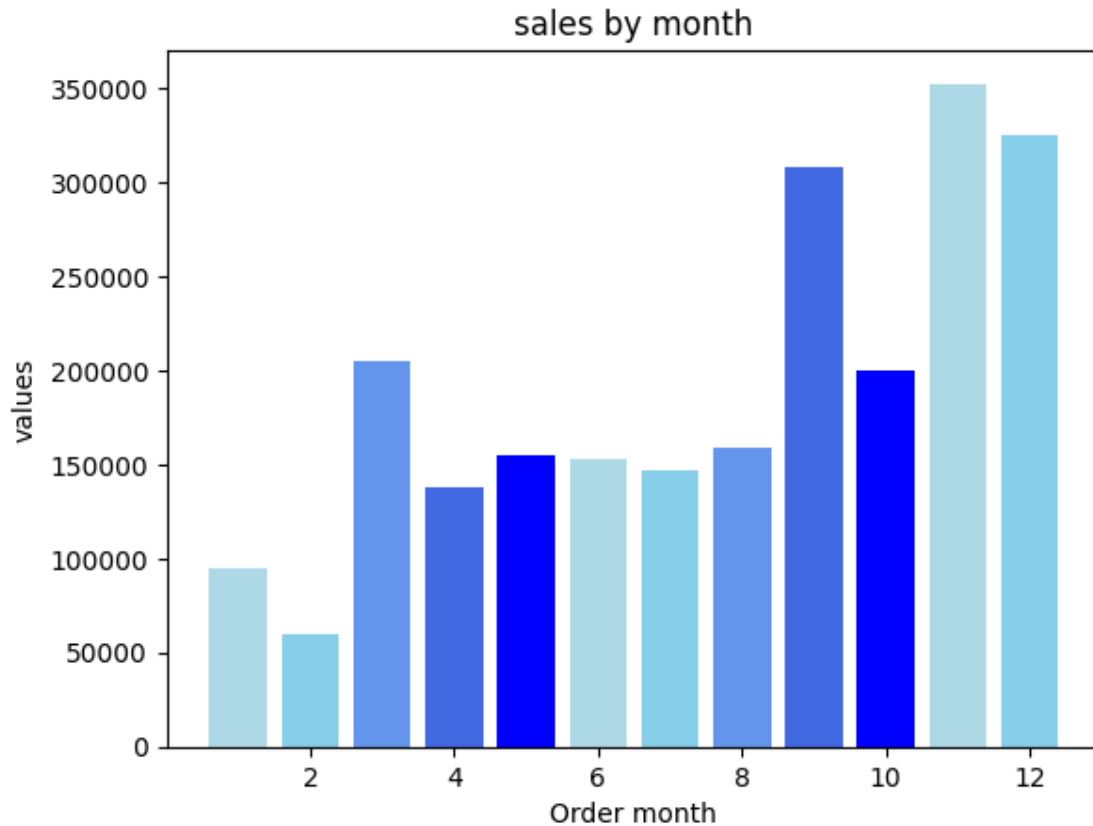
[ ]: plt.bar(sales_by_sub_category['Sub-Category'],sales_by_sub_category['Sales'])
      plt.title('Sales_of_sub_category')
      plt.xlabel('Sub-Category')
      plt.ylabel('values')
      plt.xticks(rotation=90)
      plt.show()

```



now let's have a look at the sales by month

```
[ ]: colors = ['#ADD8E6', '#87CEEB', '#6495ED', '#4169E1', '#0000FF']
plt.bar(sales_by_month['Order Month'], sales_by_month['Sales'], color=colors)
plt.title('sales by month')
plt.xlabel('Order month')
plt.ylabel('values')
plt.show()
```



```
[ ]: profit_by_month = data.groupby('Order Month')['Profit'].sum().reset_index()
profit_by_month
```

```
[ ]:
Order Month    Profit
0           1  9134.4461
1           2 10294.6107
2           3 28594.6872
3           4 11587.4363
4           5 22411.3078
5           6 21285.7954
6           7 13832.6648
7           8 21776.9384
8           9 36857.4753
9          10 31784.0413
10          11 35468.4265
11          12 43369.1919
```

now let's have a look at the profit by month using line chart



```
[ ]: plt.plot(profit_by_month['Order Month'],profit_by_month['Profit'])
plt.xlabel('Order Month')
plt.ylabel('Profit')
plt.title('Profit by month')
plt.show()
```



now let's have a look at the profit by month

```
[ ]: plt.bar(profit_by_month['Order Month'],profit_by_month['Profit'])
plt.xlabel('Order Month')
plt.ylabel('Profit')
plt.title('Profit by month')
plt.show()
```



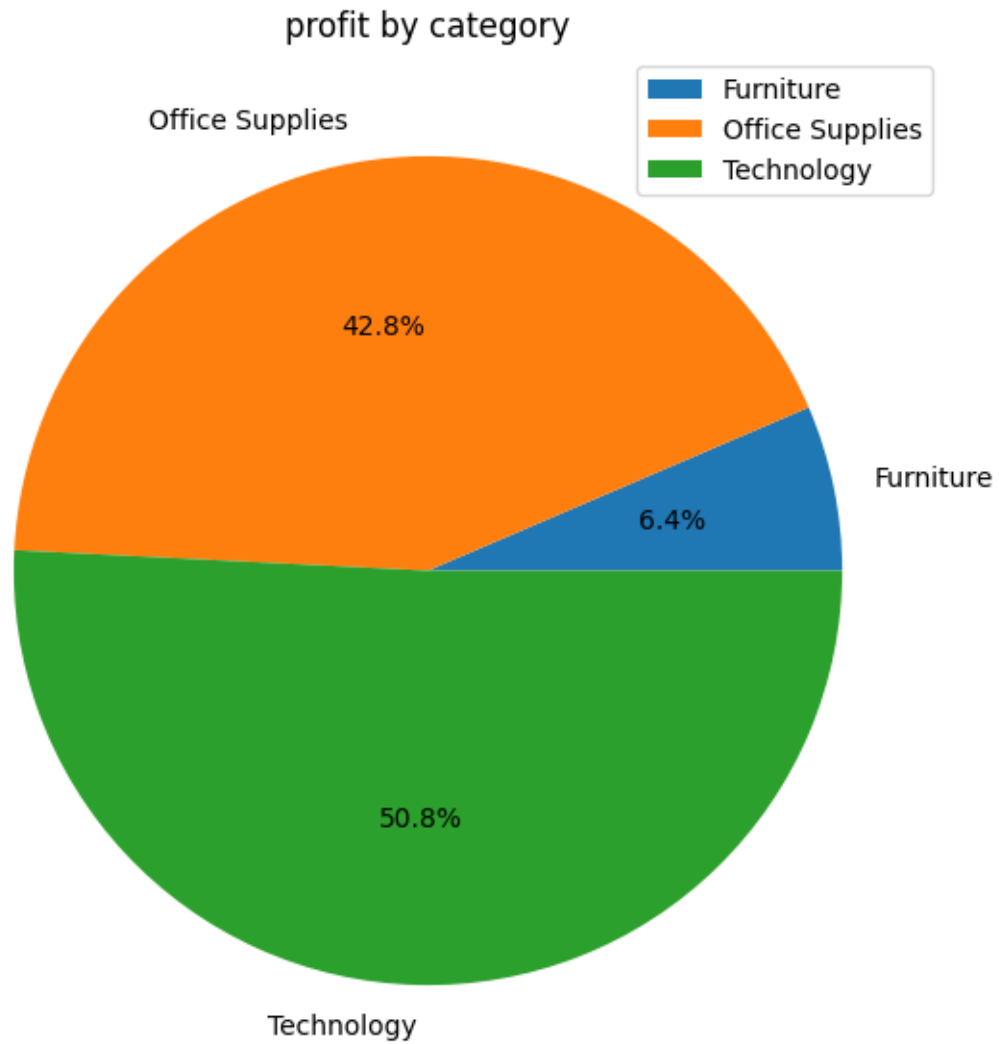
```
[ ]: profit_by_category = data.groupby('Category')['Profit'].sum().reset_index()

profit_by_category
```

```
[ ]:
      Category      Profit
0    Furniture  18451.2728
1  Office Supplies 122490.8008
2    Technology 145454.9481
```

Now let's have a look at the profit by category:

```
[ ]: labels = profit_by_category['Category']
      sizes = profit_by_category['Profit']
      plt.figure(figsize=(10,7))
      plt.pie(sizes,labels=labels,autopct='%1.1f%%')
      plt.title('profit by category')
      plt.legend()
      plt.show()
```

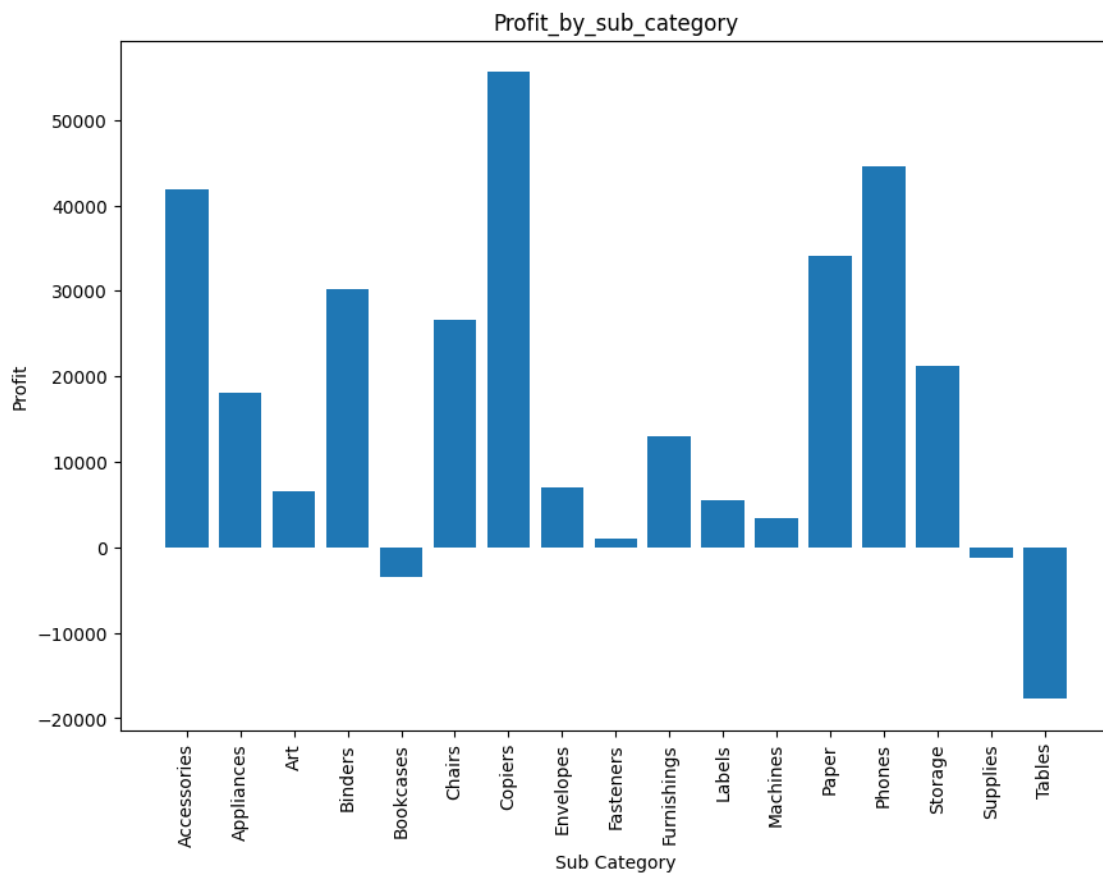


```
[ ]: profit_by_sub_category = data.groupby('Sub-Category')['Profit'].sum().  
      ↪reset_index()  
profit_by_sub_category
```

```
[ ]:   Sub-Category    Profit  
0   Accessories  41936.6357  
1   Appliances  18138.0054  
2     Art       6527.7870  
3   Binders     30221.7633  
4   Bookcases  -3472.5560  
5     Chairs   26590.1663  
6   Copiers    55617.8249  
7   Envelopes   6964.1767
```

8	Fasteners	949.5182
9	Furnishings	13059.1436
10	Labels	5546.2540
11	Machines	3384.7569
12	Paper	34053.5693
13	Phones	44515.7306
14	Storage	21278.8264
15	Supplies	-1189.0995
16	Tables	-17725.4811

```
[ ]: plt.figure(figsize=(10,7))
plt.bar(profit_by_sub_category['Sub-Category'],profit_by_sub_category['Profit'])
plt.title('Profit_by_sub_category')
plt.xticks(rotation=90)
plt.xlabel('Sub Category')
plt.ylabel('Profit')
plt.show()
```



```
[ ]: pd.set_option('display.max_columns', None)
```

```
[ ]: data.head(2)
```

```
[ ]:
  Row ID      Order ID Order Date  Ship Date      Ship Mode Customer ID \
0      1  CA-2016-152156 2016-11-08 2016-11-11  Second Class    CG-12520
1      2  CA-2016-152156 2016-11-08 2016-11-11  Second Class    CG-12520

  Customer Name  Segment      Country      City      State  Postal Code \
0  Claire Gute  Consumer  United States  Henderson  Kentucky    42420
1  Claire Gute  Consumer  United States  Henderson  Kentucky    42420

  Region      Product ID  Category Sub-Category \
0  South  FUR-BO-10001798  Furniture  Bookcases
1  South  FUR-CH-10000454  Furniture  Chairs

                                Product Name  Sales  Quantity \
0                                Bush Somerset Collection Bookcase  261.96      2
1  Hon Deluxe Fabric Upholstered Stacking Chairs,...  731.94      3

  Discount  Profit  Order Month  Order Year  Order Day of Week
0      0.0  41.9136           11        2016              1
1      0.0  219.5820           11        2016              1
```

```
[ ]: data['Segment'].value_counts()
```

```
[ ]: Consumer      5191
      Corporate     3020
      Home Office   1783
      Name: Segment, dtype: int64
```

```
[ ]: sales_profit_by_segment = data.groupby('Segment').agg({'Sales': 'sum', 'Profit':
↳ 'sum'}).reset_index()

sales_profit_by_segment
```

```
[ ]:
  Segment      Sales      Profit
0  Consumer  1.161401e+06  134119.2092
1  Corporate  7.061464e+05   91979.1340
2  Home Office 4.296531e+05   60298.6785
```

So the store has higher profits from the product sales for consumers, but the profit from corporate product sales is better in the sales-to-profit ratio. Let's have a look at it to validate our findings:

```
[ ]: sales_profit_by_segment = data.groupby('Segment').agg({'Sales': 'sum', 'Profit':
↳ 'sum'}).reset_index()
sales_profit_by_segment['Sales_to_Profit_Ratio'] =
↳ sales_profit_by_segment['Sales'] / sales_profit_by_segment['Profit']
print(sales_profit_by_segment[['Segment', 'Sales_to_Profit_Ratio']])
```

	Segment	Sales_to_Profit_Ratio
0	Consumer	8.659471
1	Corporate	7.677245
2	Home Office	7.125416

## Summary

Store sales and profit analysis help businesses identify areas for improvement and make data-driven decisions to optimize their operations, pricing, marketing, and inventory management strategies to drive revenue and growth. I hope you liked this article on the task of analyzing the sales and profit of a store using Python.

[ ]: