

In [2]:

```
import pandas as pd
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"
```

In [3]: `data = pd.read_csv("Superstore.csv",encoding = 'latin-1')`In [4]: `data.head()`

Out[4]:

	Row ID	Order ID	Order Date	Ship Date	Ship Mode	Customer ID	Customer Name	Segment	Country
0	1	CA-2016-152156	11/8/2016	11/11/2016	Second Class	CG-12520	Claire Gute	Consumer	United States
1	2	CA-2016-152156	11/8/2016	11/11/2016	Second Class	CG-12520	Claire Gute	Consumer	United States
2	3	CA-2016-138688	6/12/2016	6/16/2016	Second Class	DV-13045	Darrin Van Huff	Corporate	United States
3	4	US-2015-108966	10/11/2015	10/18/2015	Standard Class	SO-20335	Sean O'Donnell	Consumer	United States
4	5	US-2015-108966	10/11/2015	10/18/2015	Standard Class	SO-20335	Sean O'Donnell	Consumer	United States

5 rows × 21 columns

In [5]: `data.describe()`

Out[5]:

	Row ID	Postal Code	Sales	Quantity	Discount	Profit
count	9994.000000	9994.000000	9994.000000	9994.000000	9994.000000	9994.000000
mean	4997.500000	55190.379428	229.858001	3.789574	0.156203	28.656896
std	2885.163629	32063.693350	623.245101	2.225110	0.206452	234.260108
min	1.000000	1040.000000	0.444000	1.000000	0.000000	-6599.978000
25%	2499.250000	23223.000000	17.280000	2.000000	0.000000	1.728750
50%	4997.500000	56430.500000	54.490000	3.000000	0.200000	8.666500
75%	7495.750000	90008.000000	209.940000	5.000000	0.200000	29.364000
max	9994.000000	99301.000000	22638.480000	14.000000	0.800000	8399.976000

In [6]: `data.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9994 entries, 0 to 9993
Data columns (total 21 columns):
#   Column                Non-Null Count  Dtype  
---  -
0   Row ID                9994 non-null  int64  
1   Order ID              9994 non-null  object  
2   Order Date            9994 non-null  object  
3   Ship Date              9994 non-null  object  
4   Ship Mode              9994 non-null  object  
5   Customer ID            9994 non-null  object  
6   Customer Name          9994 non-null  object  
7   Segment                9994 non-null  object  
8   Country                9994 non-null  object  
9   City                   9994 non-null  object  
10  State                  9994 non-null  object  
11  Postal Code            9994 non-null  int64  
12  Region                 9994 non-null  object  
13  Product ID             9994 non-null  object  
14  Category                9994 non-null  object  
15  Sub-Category           9994 non-null  object  
16  Product Name           9994 non-null  object  
17  Sales                  9994 non-null  float64 
18  Quantity                9994 non-null  int64  
19  Discount                9994 non-null  float64 
20  Profit                 9994 non-null  float64 
dtypes: float64(3), int64(3), object(15)
memory usage: 1.6+ MB
```

Covertin data column

In [7]: `data['Order Date'] = pd.to_datetime(data['Order Date'])`
`data['Ship Date'] = pd.to_datetime(data['Ship Date'])`

```
In [8]: data.info()
```

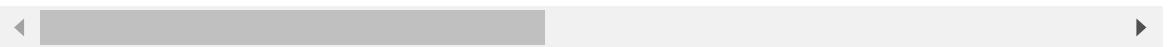
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9994 entries, 0 to 9993
Data columns (total 21 columns):
#   Column                Non-Null Count  Dtype  
---  -
0   Row ID                9994 non-null  int64   
1   Order ID              9994 non-null  object  
2   Order Date            9994 non-null  datetime64[ns]
3   Ship Date             9994 non-null  datetime64[ns]
4   Ship Mode             9994 non-null  object  
5   Customer ID           9994 non-null  object  
6   Customer Name         9994 non-null  object  
7   Segment              9994 non-null  object  
8   Country               9994 non-null  object  
9   City                 9994 non-null  object  
10  State                9994 non-null  object  
11  Postal Code          9994 non-null  int64   
12  Region              9994 non-null  object  
13  Product ID           9994 non-null  object  
14  Category             9994 non-null  object  
15  Sub-Category         9994 non-null  object  
16  Product Name         9994 non-null  object  
17  Sales                9994 non-null  float64  
18  Quantity             9994 non-null  int64   
19  Discount             9994 non-null  float64  
20  Profit              9994 non-null  float64  
dtypes: datetime64[ns](2), float64(3), int64(3), object(13)
memory usage: 1.6+ MB
```

```
In [9]: data.head()
```

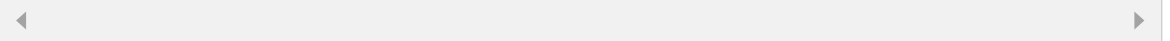
```
Out[9]:
```

	Row ID	Order ID	Order Date	Ship Date	Ship Mode	Customer ID	Customer Name	Segment	Country	Cit
0	1	CA-2016-152156	2016-11-08	2016-11-11	Second Class	CG-12520	Claire Gute	Consumer	United States	Henderso
1	2	CA-2016-152156	2016-11-08	2016-11-11	Second Class	CG-12520	Claire Gute	Consumer	United States	Henderso
2	3	CA-2016-138688	2016-06-12	2016-06-16	Second Class	DV-13045	Darrin Van Huff	Corporate	United States	Lo Angele
3	4	US-2015-108966	2015-10-11	2015-10-18	Standard Class	SO-20335	Sean O'Donnell	Consumer	United States	Fo Lauderdale
4	5	US-2015-108966	2015-10-11	2015-10-18	Standard Class	SO-20335	Sean O'Donnell	Consumer	United States	Fo Lauderdale

5 rows × 21 columns



```
In [10]: 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
```



```
In [11]: data.head()
```

```
Out[11]:
```

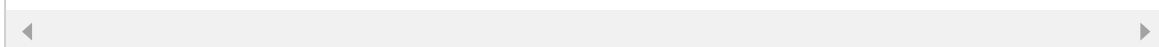
	Row ID	Order ID	Order Date	Ship Date	Ship Mode	Customer ID	Customer Name	Segment	Country	Cit
0	1	CA-2016-152156	2016-11-08	2016-11-11	Second Class	CG-12520	Claire Gute	Consumer	United States	Henderso
1	2	CA-2016-152156	2016-11-08	2016-11-11	Second Class	CG-12520	Claire Gute	Consumer	United States	Henderso
2	3	CA-2016-138688	2016-06-12	2016-06-16	Second Class	DV-13045	Darrin Van Huff	Corporate	United States	Lo Angele
3	4	US-2015-108966	2015-10-11	2015-10-18	Standard Class	SO-20335	Sean O'Donnell	Consumer	United States	Fo Lauderdale
4	5	US-2015-108966	2015-10-11	2015-10-18	Standard Class	SO-20335	Sean O'Donnell	Consumer	United States	Fo Lauderdale

5 rows × 24 columns



Monthly sales analysis

```
In [12]: sales_by_month = data.groupby('Order Month')['Sales'].sum().reset_index()
```



In [13]: sales_by_month

Out[13]:

	Order Month	Sales
0	1	94924.8356
1	2	59751.2514
2	3	205005.4888
3	4	137762.1286
4	5	155028.8117
5	6	152718.6793
6	7	147238.0970
7	8	159044.0630
8	9	307649.9457
9	10	200322.9847
10	11	352461.0710
11	12	325293.5035

```
In [14]: fig = px.line(sales_by_month,
                      x = 'Order Month',
                      y = 'Sales',
                      title = 'Monthly Sales Analysis')
fig.show()
```

Monthly Sales Analysis



sales by per category

```
In [15]: sales_by_category = data.groupby('Category')['Sales'].sum().reset_index()
```

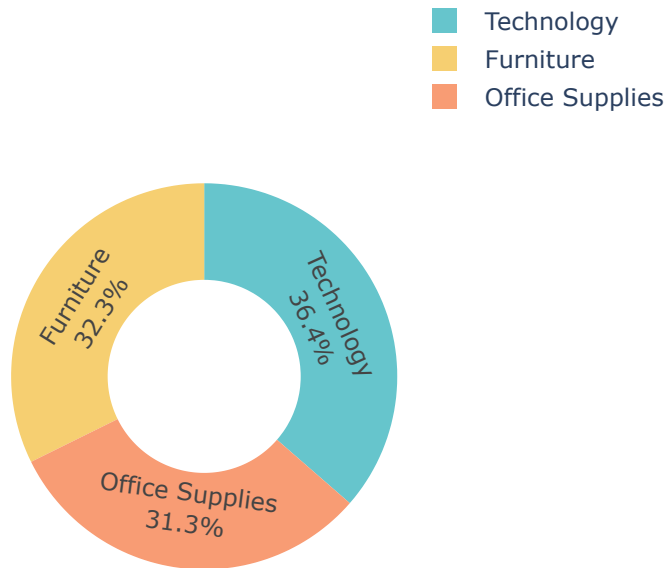
```
In [16]: sales_by_category
```

Out[16]:

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

```
In [17]: fig = px.pie(sales_by_category,
                    values = 'Sales',
                    names = 'Category',
                    hole = 0.5,
                    color_discrete_sequence = px.colors.qualitative.Pastel)
fig.update_traces(textposition = 'inside', textinfo = 'percent+label')
fig.update_layout(title_text = "Sales Analysis by Category", title_font = dict(size=18))
fig.show()
```

Sales Analysis by Category



sales analysis by subcategory

```
In [18]: sales_by_subcategory = data.groupby('Sub-Category')['Sales'].sum().reset_index()
```

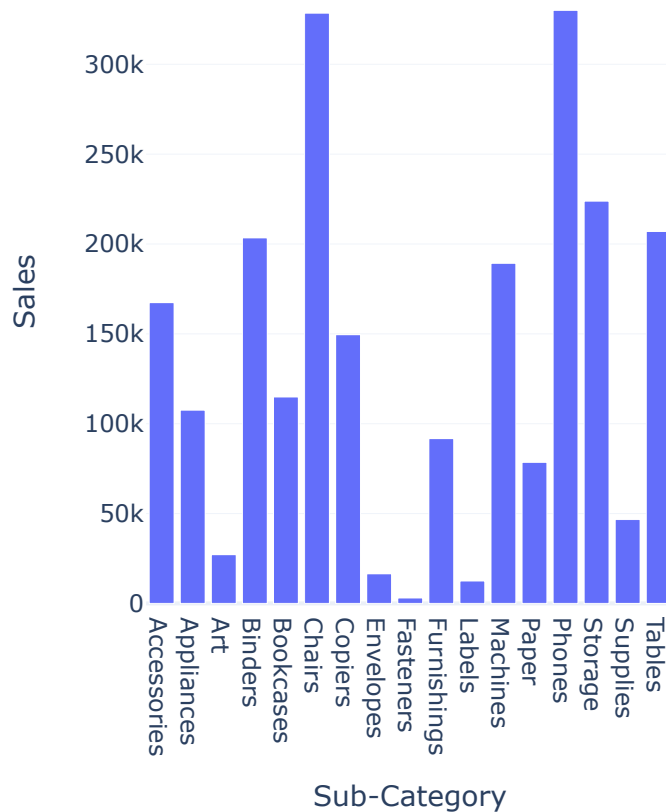

In [19]: sales_by_subcategory

Out[19]:

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

```
In [20]: fig = px.bar(sales_by_subcategory, x = 'Sub-Category', y = 'Sales', title = "  
fig.show()
```

sales by subcategory



Monthly profit analysis

```
In [21]: profit_by_month = data.groupby("Order Month")['Profit'].sum().reset_index()
```

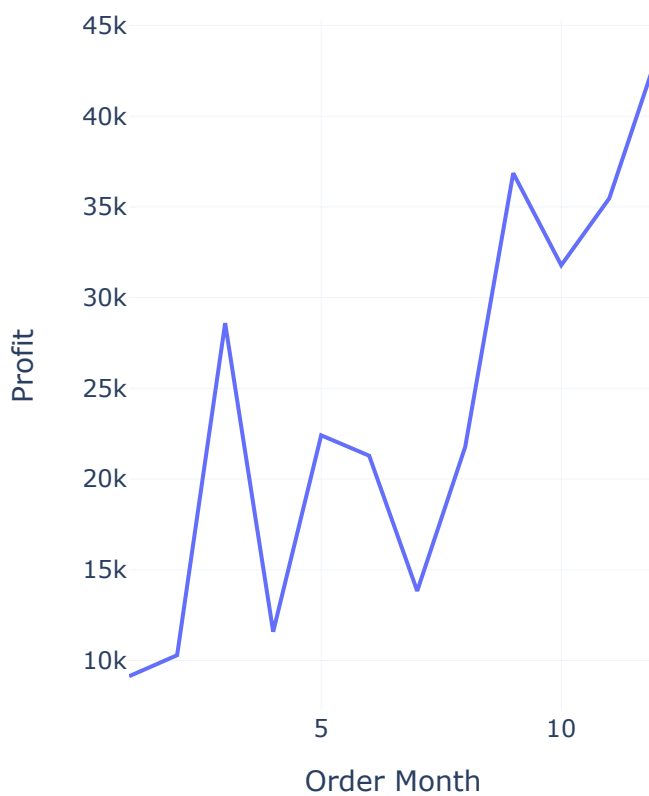
```
In [22]: profit_by_month
```

```
Out[22]:
```

	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

```
In [23]: fig = px.line(profit_by_month,x = 'Order Month', y = 'Profit',title = "Mont  
fig.show()
```

Monthly Profit Analysis



profit_by_category

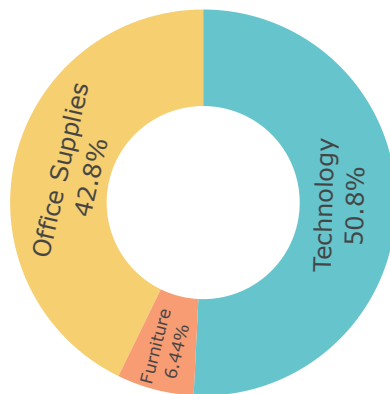
```
In [24]: profit_by_category = data.groupby('Category')['Profit'].sum().reset_index()  
profit_by_category
```

Out[24]:

	Category	Profit
0	Furniture	18451.2728
1	Office Supplies	122490.8008
2	Technology	145454.9481

```
In [43]: px.pie(profit_by_category,  
               values = 'Profit',  
               names = 'Category',  
               hole = 0.5,  
               color_discrete_sequence = px.colors.qualitative.Pastel)  
fig.update_traces(textposition = 'inside', textinfo = 'percent+label')  
fig.update_layout(title_text = "Profit by Category",title_font = dict(size = 24))  
fig.show()
```

Profit by Category



Profit by sub-category

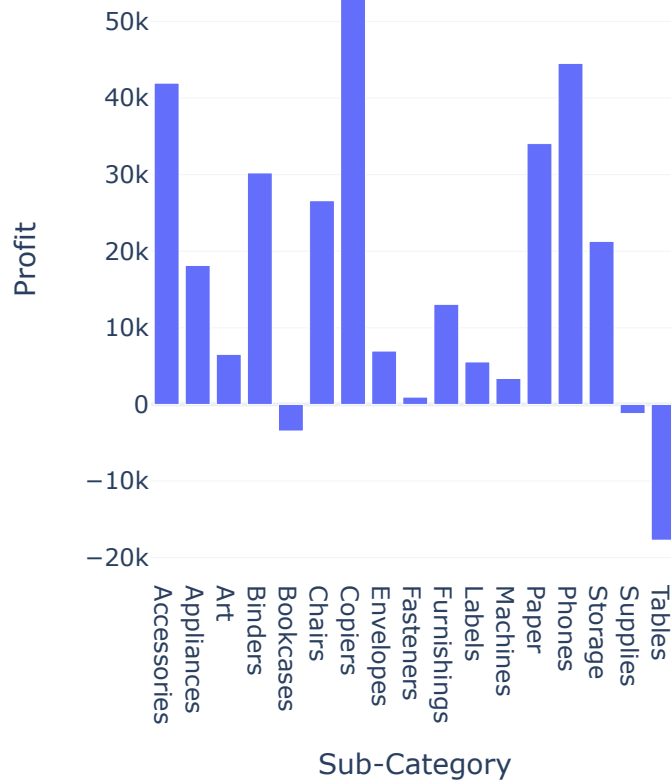
```
In [26]: profit_by_subcategory = data.groupby('Sub-Category')['Profit'].sum().reset_index()
profit_by_subcategory
```

Out[26]:

	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

```
In [27]: fig = px.bar(profit_by_subcategory, x = 'Sub-Category',y = 'Profit',title =  
fig.show()
```

Profit by subcategory



sales and profit

```
In [41]: plotly.graph_objects as go
otly.colors import qualitative as colors

gating sales and profit by segment
rofit_by_segment = data.groupby('Segment').agg({'Sales': 'sum', 'Profit': 's

ing color palette
alette = colors.Pastel

ing the figure
o.Figure()

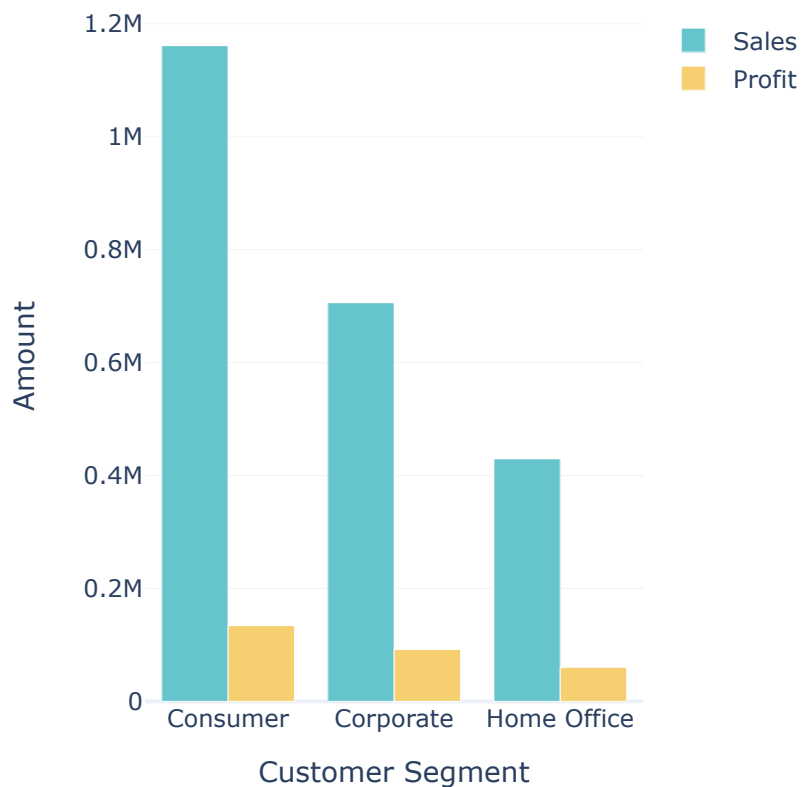
g sales bar trace
_trace(go.Bar(
ales_profit_by_segment['Segment'],
ales_profit_by_segment['Sales'],
e='Sales',
ker_color=color_palette[0]

g profit bar trace
_trace(go.Bar(
ales_profit_by_segment['Segment'],
ales_profit_by_segment['Profit'],
e='Profit',
ker_color=color_palette[1]

ing layout
ate_layout(
le='Sales and Profit Analysis by Customer Segment',
is_title='Customer Segment',
is_title='Amount',
mode='group' # Ensures bars are grouped side-by-side

ay the plot
w()
```

Sales and Profit Analysis by Customer Segment



sales to profit ratio

```
In [42]: sales_profit_by_segment = data.groupby('Segment').agg({'Sales': 'sum', 'Profit': 'sum'})
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

conclusion


```
In [ ]: 1.In the january sales less and Nov has highest sales  
        2. Technolgy> furniture>office items selles  
        3.By subcategory phone seles high  
        4. Dec profit Hlgh and jan very low  
        5. Highest profit in technology and in subcategory copiers  
        6. consumer > corporate > home office buy anything  
        7. consumer ratio is high
```