

Project Overview

This project analyzes sales data from Dmart, focusing on key business metrics like sales, profit, discounts, and customer behavior. The dataset contains 8,047 records with 17 attributes, including customer details, product categories, order information, and financials. The analysis covers data cleaning, outlier detection, exploratory data analysis (EDA), and visualizations to uncover trends and patterns.

Objective

To analyze sales performance, profitability, and discount impact while identifying key trends, top customers, and optimization opportunities.

```
#importing necessary libraries
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
import plotly.express as px
```

Data Loading

```
df = pd.read_excel("C:/Users/KIIT/Desktop/Projects/DMart Sales/DMart
Data Store.xlsx")

# to preview the data
df
```

	Order ID	Order Date	Customer Name	Country	\
0	BN-2011-7407039	2011-01-01	Ruby Patel	Sweden	
1	AZ-2011-9050313	2011-01-03	Summer Hayward	United Kingdom	
2	AZ-2011-6674300	2011-01-04	Devin Huddleston	France	
3	BN-2011-2819714	2011-01-04	Mary Parker	United Kingdom	
4	BN-2011-2819714	2011-01-04	Mary Parker	United Kingdom	
...
8042	AZ-2014-766953	2014-12-31	Jose Gambino	United Kingdom	
8043	BN-2014-4140795	2014-12-31	Daniel Hamilton	Netherlands	
8044	BN-2014-4140795	2014-12-31	Daniel Hamilton	Netherlands	
8045	BN-2014-4140795	2014-12-31	Daniel Hamilton	Netherlands	
8046	AZ-2014-766953	2014-12-31	Jose Gambino	United Kingdom	

	State	City	Region	Segment	Ship
Mode \					
0	Stockholm	Stockholm	North	Home Office	Economy
Plus					
1	England	Southport	North	Consumer	
Economy					
2	Auvergne-Rhône-Alpes	Valence	Central	Consumer	
Economy					
3	England	Birmingham	North	Corporate	

Economy						
4	England	Birmingham	North	Corporate		
Economy						
...		
...						
8042	England	Maidenhead	North	Corporate		
Economy						
8043	North Brabant	Eindhoven	Central	Home Office	Economy	
Plus						
8044	North Brabant	Eindhoven	Central	Home Office	Economy	
Plus						
8045	North Brabant	Eindhoven	Central	Home Office	Economy	
Plus						
8046	England	Maidenhead	North	Corporate		
Economy						

	Category	Sub-Category	Product
Name \			
0	Office Supplies	Paper	Enermax Note Cards, Premium
1	Furniture	Bookcases	Dania Corner Shelving, Traditional
2	Office Supplies	Art	Binney & Smith Sketch Pad, Easy-Erase
3	Office Supplies	Art	Boston Markers, Easy-Erase
4	Office Supplies	Storage	Eldon Folders, Single Width
...	
...			
8042	Furniture	Bookcases	Ikea Stackable Bookrack, Traditional
8043	Office Supplies	Art	BIC Pencil Sharpener, Fluorescent
8044	Office Supplies	Binders	Avery Binder Covers, Recycled
8045	Technology	Machines	StarTech Phone, Red
8046	Technology	Phones	Motorola Audio Dock, VoIP

	Discount	Sales	Profit	Quantity	Feedback?
0	0.5	45	-26	3	False
1	0.0	854	290	7	True
2	0.0	140	21	3	True
3	0.5	27	-22	2	True
4	0.5	17	-1	2	True
...
8042	0.0	245	91	2	True

8043	0.5	30	-10	2	False
8044	0.5	23	-6	4	True
8045	0.5	108	-19	3	False
8046	0.0	867	251	5	False

[8047 rows x 17 columns]

#to check for the first 10 rows
df.head(10)

	Order ID	Order Date	Customer Name	Country	\
0	BN-2011-7407039	2011-01-01	Ruby Patel	Sweden	
1	AZ-2011-9050313	2011-01-03	Summer Hayward	United Kingdom	
2	AZ-2011-6674300	2011-01-04	Devin Huddleston	France	
3	BN-2011-2819714	2011-01-04	Mary Parker	United Kingdom	
4	BN-2011-2819714	2011-01-04	Mary Parker	United Kingdom	
5	AZ-2011-617423	2011-01-05	Daniel Burke	France	
6	AZ-2011-617423	2011-01-05	Daniel Burke	France	
7	AZ-2011-2918397	2011-01-07	Fredrick Beveridge	France	
8	AZ-2011-2918397	2011-01-07	Fredrick Beveridge	France	
9	AZ-2011-2918397	2011-01-07	Fredrick Beveridge	France	

	State	City	Region	Segment
0	Stockholm	Stockholm	North	Home Office
1	England	Southport	North	Consumer
2	Auvergne-Rhône-Alpes	Valence	Central	Consumer
3	England	Birmingham	North	Corporate
4	England	Birmingham	North	Corporate
5	Auvergne-Rhône-Alpes	Echirolles	Central	Home Office
6	Auvergne-Rhône-Alpes	Echirolles	Central	Home Office
7	Provence-Alpes-Côte d'Azur	La Seyne-sur-Mer	Central	Corporate
8	Provence-Alpes-Côte d'Azur	La Seyne-sur-Mer	Central	Corporate
9	Provence-Alpes-Côte d'Azur	La Seyne-sur-Mer	Central	Corporate

	Ship Mode	Category	Sub-Category	\
0	Economy Plus	Office Supplies	Paper	
1	Economy	Furniture	Bookcases	
2	Economy	Office Supplies	Art	
3	Economy	Office Supplies	Art	
4	Economy	Office Supplies	Storage	

5	Priority	Office Supplies	Art
6	Priority	Office Supplies	Art
7	Priority	Furniture	Bookcases
8	Priority	Office Supplies	Fasteners
9	Priority	Office Supplies	Storage

	Product Name	Discount	Sales	Profit \
0	Enermax Note Cards, Premium	0.5	45	-
26				
1	Dania Corner Shelving, Traditional	0.0	854	
290				
2	Binney & Smith Sketch Pad, Easy-Erase	0.0	140	
21				
3	Boston Markers, Easy-Erase	0.5	27	-
22				
4	Eldon Folders, Single Width	0.5	17	
-1				
5	Binney & Smith Pencil Sharpener, Water Color	0.0	90	
21				
6	Sanford Canvas, Fluorescent	0.0	207	
77				
7	Bush Floating Shelf Set, Pine	0.1	155	
36				
8	Accos Thumb Tacks, Assorted Sizes	0.0	33	
2				
9	Smead Lockers, Industrial	0.1	716	
143				

	Quantity	Feedback?
0	3	False
1	7	True
2	3	True
3	2	True
4	2	True
5	3	False
6	4	False
7	1	True
8	3	True
9	4	True

#to check for the last 10 rows
df.tail(10)

	Order ID	Order Date	Customer Name	Country
\				
8037	AZ-2014-436448	2014-12-30	Georgia Arundale	Italy
8038	AZ-2014-1412225	2014-12-31	Leon Barnes	United Kingdom

8039	AZ-2014-4217323	2014-12-31	Evie Morton	France
8040	AZ-2014-8174835	2014-12-31	Eloise Sykes	Germany
8041	AZ-2014-7604524	2014-12-31	Rebecca Chamberlain	Germany
8042	AZ-2014-766953	2014-12-31	Jose Gambino	United Kingdom
8043	BN-2014-4140795	2014-12-31	Daniel Hamilton	Netherlands
8044	BN-2014-4140795	2014-12-31	Daniel Hamilton	Netherlands
8045	BN-2014-4140795	2014-12-31	Daniel Hamilton	Netherlands
8046	AZ-2014-766953	2014-12-31	Jose Gambino	United Kingdom
		State	City	Region
Ship Mode \				Segment
8037		Campania	Naples	South
Economy				Corporate
8038		England	Worcester	North
Priority				Consumer
8039		Normandy	Caen	Central
Economy Plus				Consumer
8040	North Rhine-Westphalia	Bielefeld	Central	Consumer
Economy				
8041	Hamburg	Hamburg	Central	Home Office
Economy				
8042	England	Maidenhead	North	Corporate
Economy				
8043	North Brabant	Eindhoven	Central	Home Office
Economy Plus				
8044	North Brabant	Eindhoven	Central	Home Office
Economy Plus				
8045	North Brabant	Eindhoven	Central	Home Office
Economy Plus				
8046	England	Maidenhead	North	Corporate
Economy				
		Category	Sub-Category	Product
Name \				
8037	Office Supplies	Binders		Acco Binder,
Economy				
8038	Office Supplies	Storage		Fellowes Shelving, Single
Width				
8039	Office Supplies	Storage		Fellowes Lockers, Wire
Frame				
8040	Office Supplies	Supplies		Kleencut Shears,
Serrated				

8041	Office Supplies	Binders	Wilson Jones Index Tab, Economy
8042	Furniture	Bookcases	Ikea Stackable Bookrack, Traditional
8043	Office Supplies	Art	BIC Pencil Sharpener, Fluorescent
8044	Office Supplies	Binders	Avery Binder Covers, Recycled
8045	Technology	Machines	StarTech Phone, Red
8046	Technology	Phones	Motorola Audio Dock, VoIP

	Discount	Sales	Profit	Quantity	Feedback?
8037	0.0	45	6	3	True
8038	0.0	289	75	5	True
8039	0.1	557	217	3	False
8040	0.0	261	13	6	True
8041	0.0	32	8	5	True
8042	0.0	245	91	2	True
8043	0.5	30	-10	2	False
8044	0.5	23	-6	4	True
8045	0.5	108	-19	3	False
8046	0.0	867	251	5	False

Understanding the dataset

```
#the total column and rows
df.shape

(8047, 17)

#to check column names
df.columns

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

#data types and null values
df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8047 entries, 0 to 8046
Data columns (total 17 columns):
#   Column                Non-Null Count  Dtype
---  -

```

0	Order ID	8047	non-null	object
1	Order Date	8047	non-null	datetime64[ns]
2	Customer Name	8047	non-null	object
3	Country	8047	non-null	object
4	State	8047	non-null	object
5	City	8047	non-null	object
6	Region	8047	non-null	object
7	Segment	8047	non-null	object
8	Ship Mode	8047	non-null	object
9	Category	8047	non-null	object
10	Sub-Category	8047	non-null	object
11	Product Name	8047	non-null	object
12	Discount	8047	non-null	float64
13	Sales	8047	non-null	int64
14	Profit	8047	non-null	int64
15	Quantity	8047	non-null	int64
16	Feedback?	8047	non-null	bool

dtypes: bool(1), datetime64[ns](1), float64(1), int64(3), object(11)
memory usage: 1013.9+ KB

#to check for the null values

df.isnull().sum()

Order ID	0
Order Date	0
Customer Name	0
Country	0
State	0
City	0
Region	0
Segment	0
Ship Mode	0
Category	0
Sub-Category	0
Product Name	0
Discount	0
Sales	0
Profit	0
Quantity	0
Feedback?	0

dtype: int64

#to check for duplicates

duplicate = df.duplicated().sum()

duplicate #2 duplicate rows

2

#to check for the unique values

df.nunique()

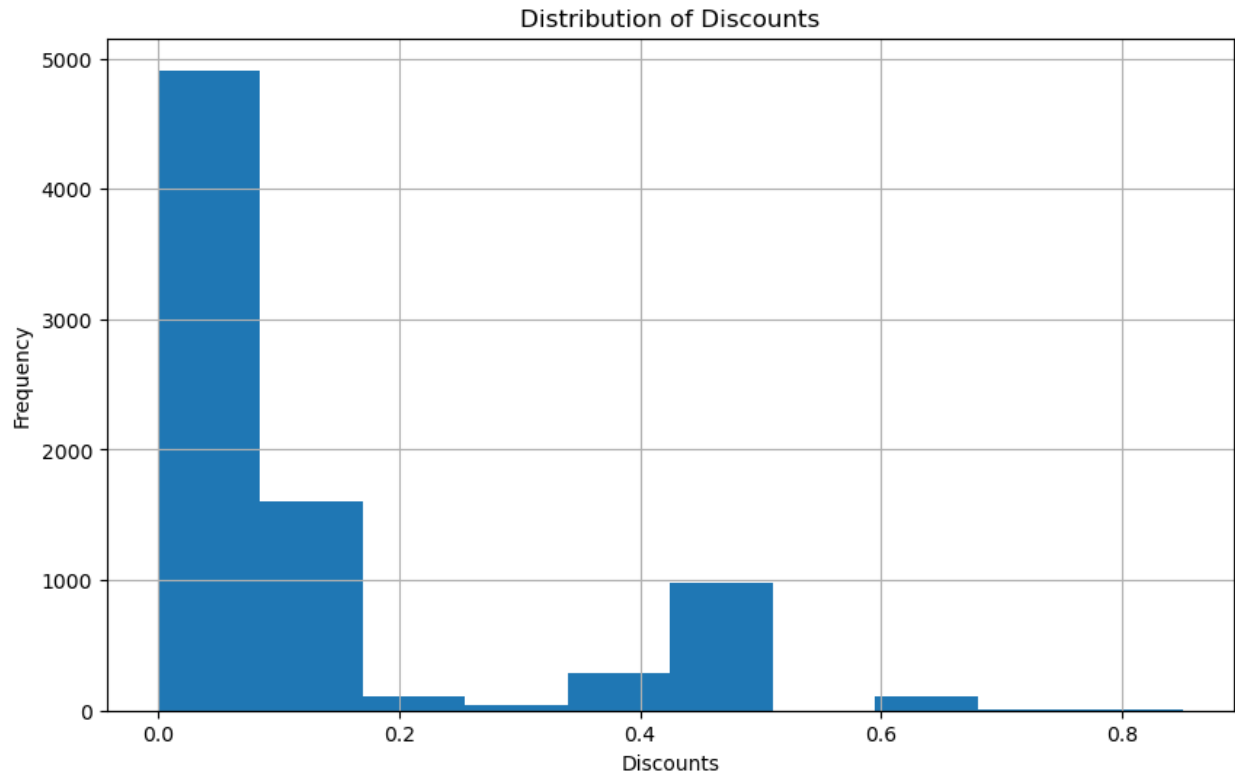
Order ID	4117
Order Date	1214
Customer Name	792
Country	15
State	127
City	999
Region	3
Segment	3
Ship Mode	4
Category	3
Sub-Category	17
Product Name	1810
Discount	14
Sales	1248
Profit	845
Quantity	14
Feedback?	2

dtype: int64

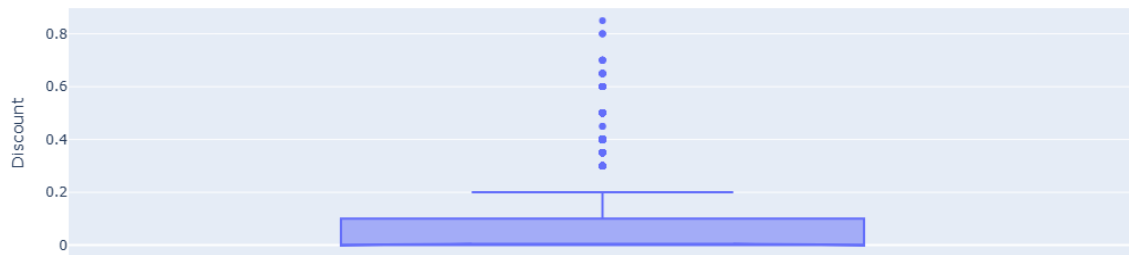
Outliers Detection and Treatment

For Discounts

```
df['Discount'].hist(bins = 10, figsize= (10,6))  
plt.title('Distribution of Discounts')  
plt.xlabel('Discounts')  
plt.ylabel('Frequency')  
plt.show()
```

```
#identifying the outliers  
fig = px.box(df, y = 'Discount', notched = True)  
fig.show()
```



```
q1 = df['Discount'].quantile(0.25)  
q2 = df['Discount'].quantile(0.5)  
q3 = df['Discount'].quantile(0.75)  
iqr = q3 - q1  
  
lb_dis = q1 - 1.5*iqr  
lb_dis  
  
-0.15000000000000002
```

```
ub_dis = q3 + 1.5*iqr
ub_dis
```

```
0.25
```

```
df[(df['Discount'] < lb_dis) | (df['Discount'] > ub_dis)]
```

	Order ID	Order Date	Customer Name	Country \
0	BN-2011-7407039	2011-01-01	Ruby Patel	Sweden
3	BN-2011-2819714	2011-01-04	Mary Parker	United Kingdom
4	BN-2011-2819714	2011-01-04	Mary Parker	United Kingdom
10	BN-2011-3248724	2011-01-08	Archer Hort	France
11	BN-2011-3248724	2011-01-08	Archer Hort	France
...
8023	BN-2014-3913645	2014-12-29	Mark Washington	Ireland
8029	BN-2014-8679573	2014-12-30	Dennis Conaway	Netherlands
8043	BN-2014-4140795	2014-12-31	Daniel Hamilton	Netherlands
8044	BN-2014-4140795	2014-12-31	Daniel Hamilton	Netherlands
8045	BN-2014-4140795	2014-12-31	Daniel Hamilton	Netherlands

	State	City	Region
Segment \			
0	Stockholm	Stockholm	North Home
Office			
3	England	Birmingham	North
Corporate			
4	England	Birmingham	North
Corporate			
10	Languedoc-Roussillon-Midi-Pyrénées	Toulouse	Central
Consumer			
11	Languedoc-Roussillon-Midi-Pyrénées	Toulouse	Central
Consumer			
...
...			
8023	Dublin	Dublin	North Home
Office			
8029	South Holland	The Hague	Central
Consumer			
8043	North Brabant	Eindhoven	Central Home
Office			
8044	North Brabant	Eindhoven	Central Home
Office			
8045	North Brabant	Eindhoven	Central Home
Office			

	Ship Mode	Category	Sub-Category \
0	Economy Plus	Office Supplies	Paper
3	Economy	Office Supplies	Art
4	Economy	Office Supplies	Storage
10	Economy	Furniture	Bookcases

11	Economy	Office Supplies	Art
...
8023	Economy	Technology	Copiers
8029	Priority	Office Supplies	Appliances
8043	Economy Plus	Office Supplies	Art
8044	Economy Plus	Office Supplies	Binders
8045	Economy Plus	Technology	Machines

Quantity \	Product Name	Discount	Sales	Profit
0	Enermax Note Cards, Premium	0.5	45	-26
3				
3	Boston Markers, Easy-Erase	0.5	27	-22
2				
4	Eldon Folders, Single Width	0.5	17	-1
2				
10	Ikea Classic Bookcase, Metal	0.6	987	-1012
6				
11	Binney & Smith Sketch Pad, Blue	0.5	116	-56
5				
...
...				
8023	Sharp Ink, Laser	0.5	373	-254
6				
8029	Cuisinart Blender, Silver	0.5	68	-62
2				
8043	BIC Pencil Sharpener, Fluorescent	0.5	30	-10
2				
8044	Avery Binder Covers, Recycled	0.5	23	-6
4				
8045	StarTech Phone, Red	0.5	108	-19
3				

	Feedback?
0	False
3	True
4	True
10	True
11	False
...	...
8023	False
8029	False
8043	False
8044	True
8045	False

[1426 rows x 17 columns]

```

#Outlier Treatment
df['Discount'] = df['Discount'].apply(lambda x: ub_dis if x > ub_dis
else (lb_dis if x < lb_dis else x))

min_dis = df['Discount'].min()
min_dis

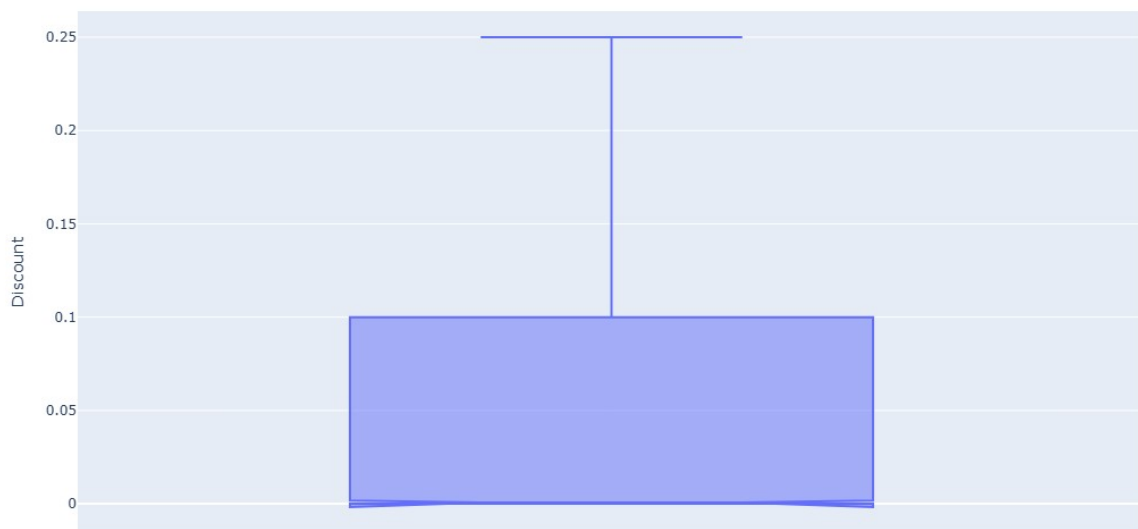
0.0

max_dis = df['Discount'].max()
max_dis

0.25

fig = px.box(df, y = 'Discount', notched = True, width=800,
height=600)
fig.show()

```

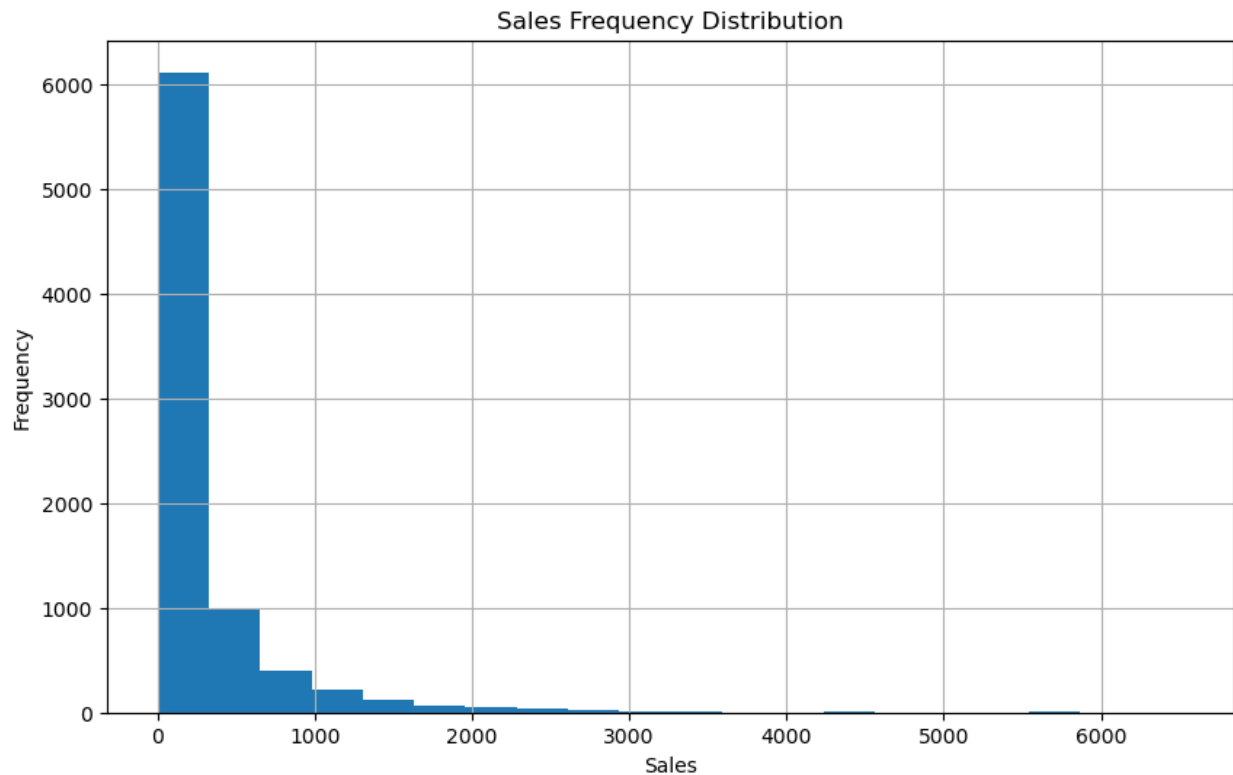


For Sales

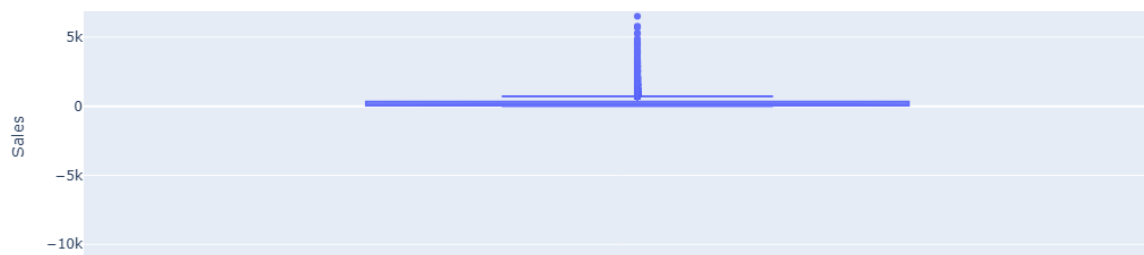
```

df['Sales'].hist(bins = 20, figsize= (10,6))
plt.title('Sales Frequency Distribution')
plt.xlabel('Sales')
plt.ylabel('Frequency')
plt.show()

```



```
#identifying the outliers  
fig = px.box(df, y = 'Sales', notched = True)  
fig.show()
```



```
q1_sales = df['Sales'].quantile(0.25)  
q2_sales = df['Sales'].quantile(0.5)  
q3_sales = df['Sales'].quantile(0.75)  
iqr_sales = q3_sales - q1_sales  
  
lb_sales = (q1_sales - 1.5 * iqr_sales)  
lb_sales  
  
-349.5
```

```

lb_sales = max(0, q1_sales - 1.5 * iqr_sales)
lb_sales

0

ub_sales = q3_sales + 1.5*iqr_sales
ub_sales

710.5

#outlier treatment
df['Sales'] = df['Sales'].apply(lambda x: ub_sales if x > ub_sales
else (lb_sales if x < lb_sales else x))

min_sales = df['Sales'].min()
min_sales

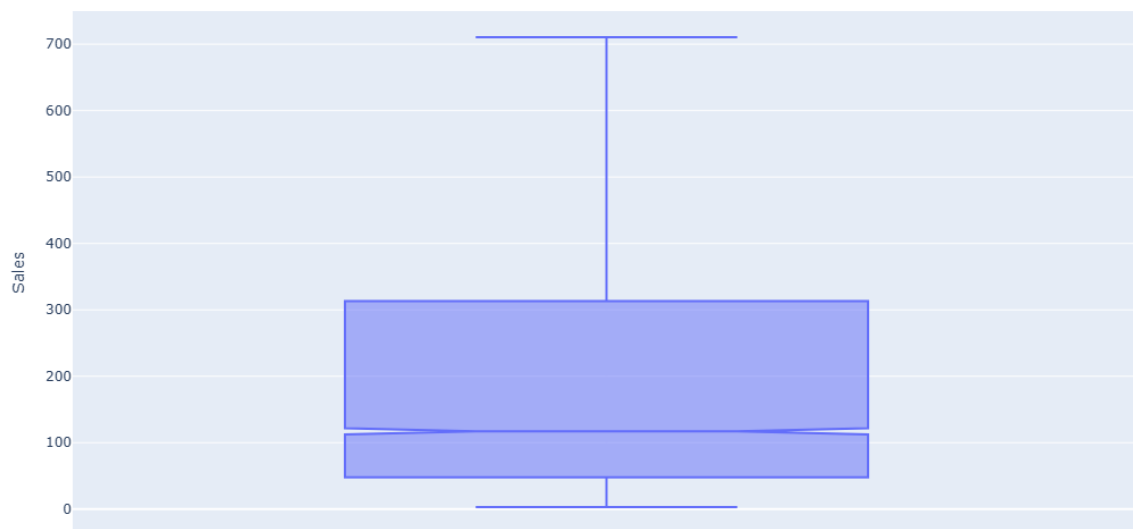
3.0

max_sales = df['Sales'].max()
max_sales

710.5

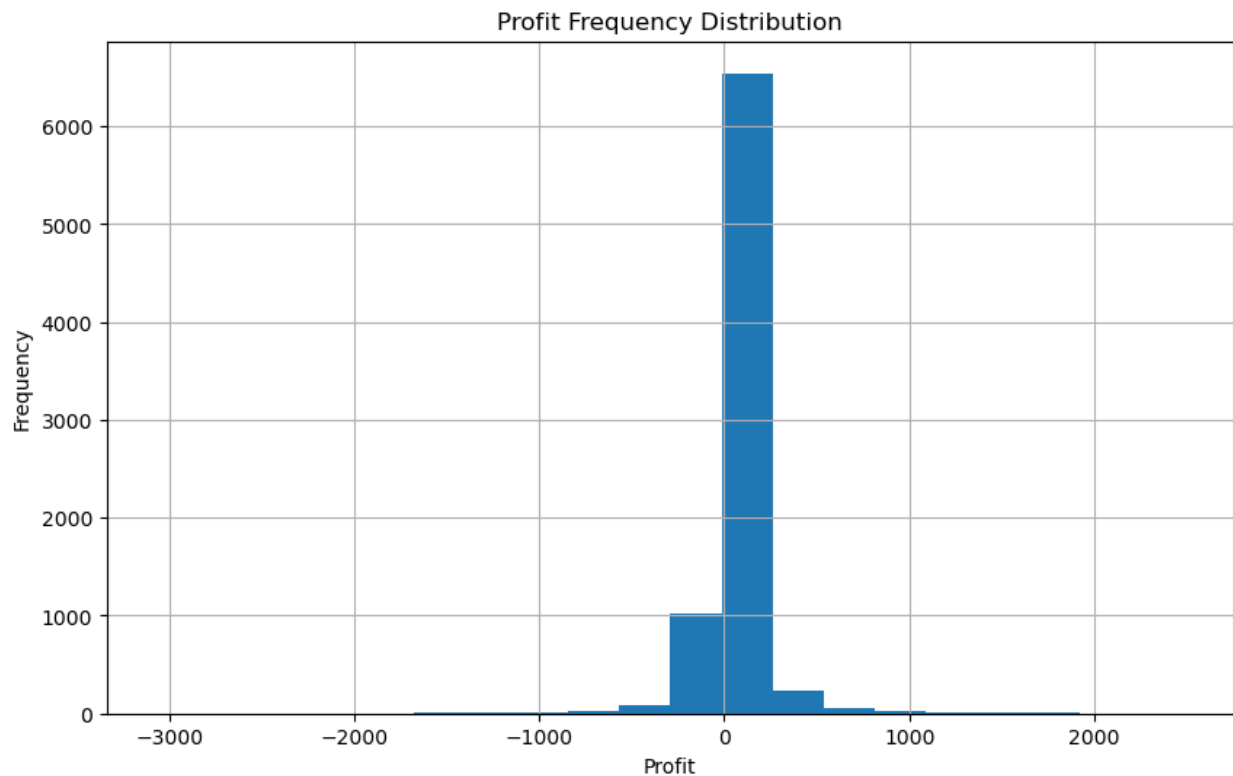
fig = px.box(df, y = 'Sales', notched = True, width=800, height=600)
fig.show()

```

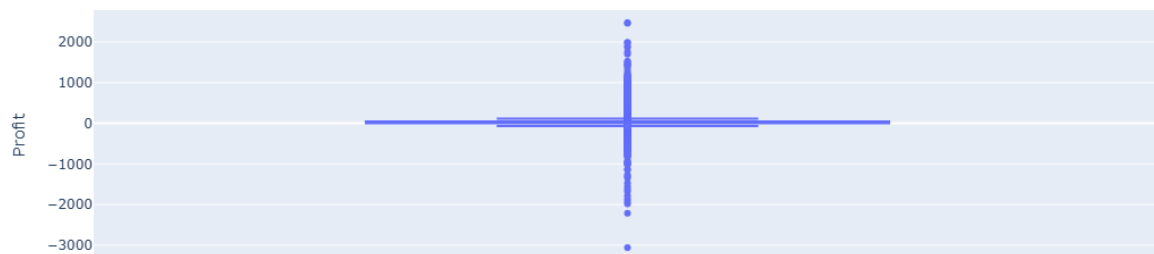


For Profit

```
df['Profit'].hist(bins = 20, figsize= (10,6))
plt.title('Profit Frequency Distribution')
plt.xlabel('Profit')
plt.ylabel('Frequency')
plt.show()
```



```
#identifying the outliers
fig = px.box(df, y = 'Profit', notched = True)
fig.show()
```



```
q1_profit = df['Profit'].quantile(0.25)
q2_profit = df['Profit'].quantile(0.5)
```

```

q3_profit = df['Profit'].quantile(0.75)
iqr_profit = q3_profit - q1_profit

lb_profit = q1_profit - 1.5*iqr_profit
lb_profit

-68.0

ub_profit = q3_profit + 1.5*iqr_profit
ub_profit

116.0

#outlier treatment
df['Profit'] = df['Profit'].apply(lambda x: ub_profit if x > ub_profit
else (lb_profit if x < lb_profit else x))

min_profit = df['Profit'].min()
min_profit

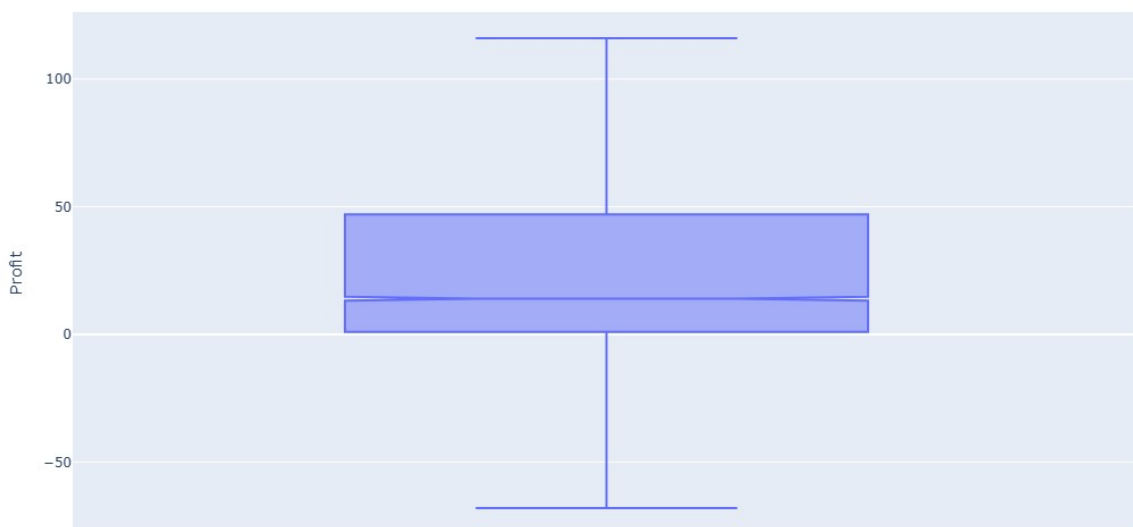
-68.0

max_profit = df['Profit'].max()
max_profit

116.0

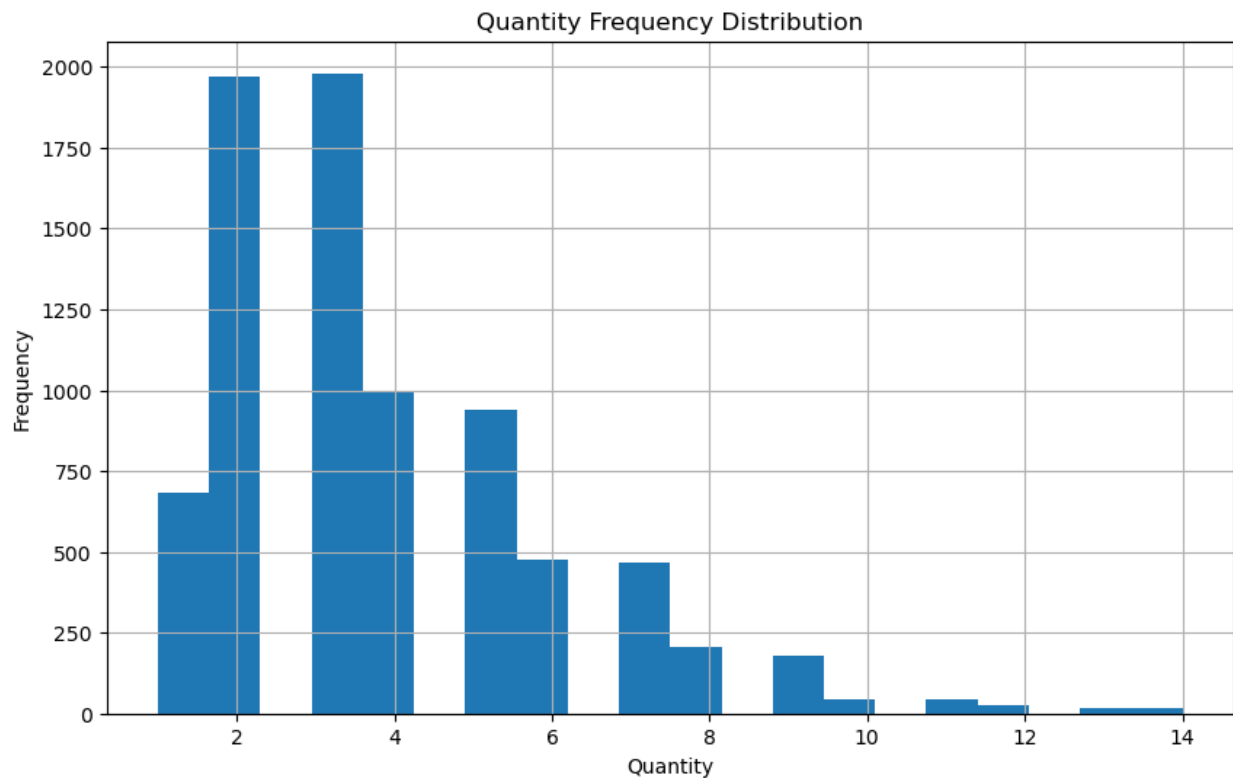
fig = px.box(df, y = 'Profit', notched = True, width=800, height=600)
fig.show()

```



For Quantity


```
df['Quantity'].hist(bins = 20, figsize= (10,6))
plt.title('Quantity Frequency Distribution')
plt.xlabel('Quantity')
plt.ylabel('Frequency')
plt.show()
```



```
#identifying the outliers
fig = px.box(df, y = 'Quantity', notched = True)
fig.show()
```



```
q1_quan = df['Quantity'].quantile(0.25)
q2_quan = df['Quantity'].quantile(0.5)
```

```

q3_quan = df['Quantity'].quantile(0.75)
iqr_quan = q3_quan - q1_quan

lb_quan = q1_quan - 1.5*iqr_quan
lb_quan

-2.5

lb_quan = max(0, q1_quan - 1.5 * iqr_quan)
lb_quan

0

ub_quan = q3_quan + 1.5*iqr_quan
ub_quan

9.5

#outlier treatment
df['Quantity'] = df['Quantity'].apply(lambda x: ub_quan if x > ub_quan
else (lb_quan if x < lb_quan else x))

min_quan = df['Quantity'].min()
min_quan

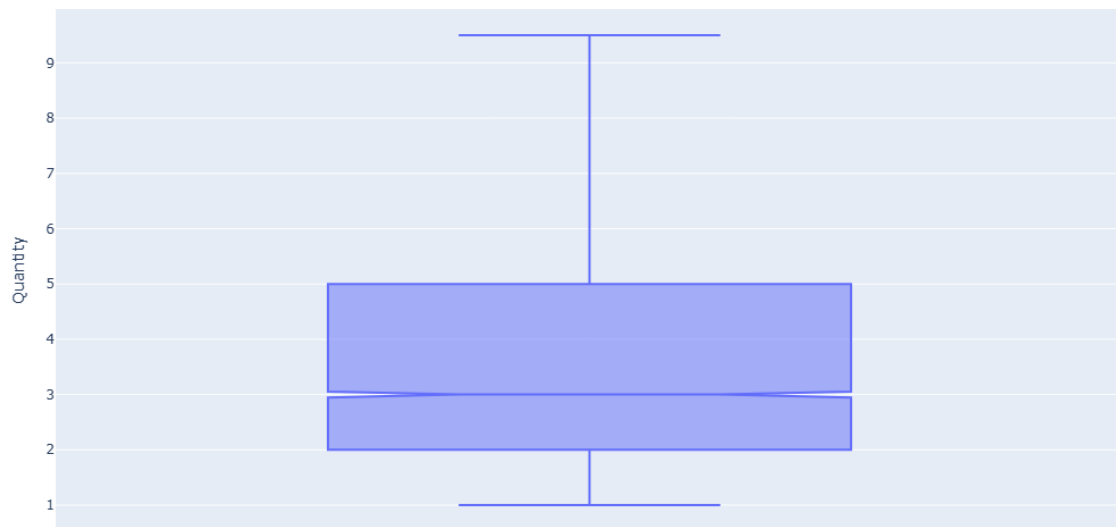
1.0

max_quan = df['Quantity'].max()
max_quan

9.5

fig = px.box(df, y = 'Quantity', notched = True, width=800,
height=600)
fig.show()

```



```
df.describe()
```

	Order Date	Discount	Sales
Profit \			
count	8047	8047.000000	8047.000000
8047.000000			
mean	2013-04-19 12:25:40.748104704	0.068727	218.702498
24.168137			
min	2011-01-01 00:00:00	0.000000	3.000000
68.000000			-
25%	2012-06-08 00:00:00	0.000000	48.000000
1.000000			
50%	2013-06-11 00:00:00	0.000000	117.000000
14.000000			
75%	2014-04-30 00:00:00	0.100000	313.000000
47.000000			
max	2014-12-31 00:00:00	0.250000	710.500000
116.000000			
std	NaN	0.096626	227.544531
48.971611			

	Quantity
count	8047.000000
mean	3.735119
min	1.000000
25%	2.000000
50%	3.000000
75%	5.000000

```
max      9.500000
std      2.079027
```

#correlation between the columns having numerical values

```
df_numeric = df.select_dtypes(include=['number'])
correlation_matrix = df_numeric.corr()
correlation_matrix
```

	Discount	Sales	Profit	Quantity
Discount	1.000000	0.052092	-0.499028	0.000088
Sales	0.052092	1.000000	0.442246	0.363168
Profit	-0.499028	0.442246	1.000000	0.172306
Quantity	0.000088	0.363168	0.172306	1.000000

#Sales & Profit Correlation

```
plt.figure(figsize=(7, 5))
correlation_matrix = df.select_dtypes(include=['number']).corr() #
Ensuring only numeric columns are used
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm',
fmt=".2f") # fmt=".2f" limits decimal places
plt.title("Sales & Profit Correlation")
plt.show()
```

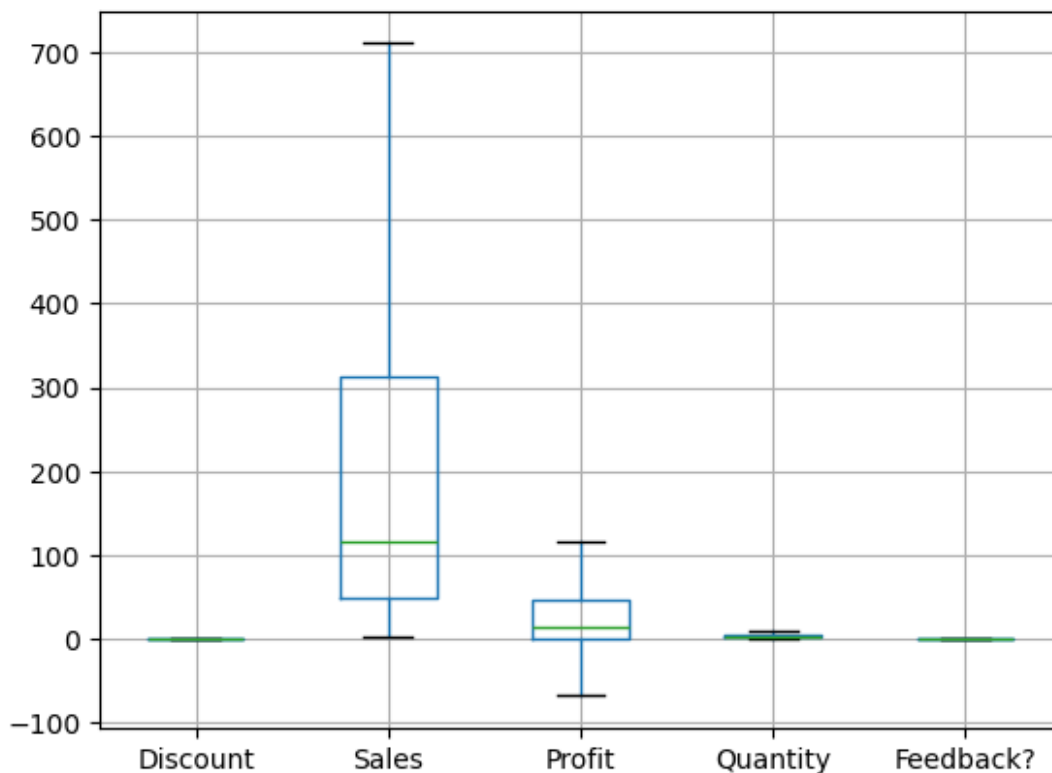


Sales and Profit are positively correlated (0.44), meaning higher sales generally lead to higher profit. Discount has a negative correlation (-0.49) with profit, meaning increasing discounts reduces profitability. Quantity and Profit have a weak correlation (0.17), indicating selling more units does not always mean higher profits.

Data Visualization

```
df.boxplot()
```

<Axes: >



```
pip install seaborn --upgrade
```

Defaulting to user installation because normal site-packages is not writeable
Note: you may need to restart the kernel to use updated packages.

Requirement already satisfied: seaborn in d:\anaconda\lib\site-packages (0.13.2)

Requirement already satisfied: numpy!=1.24.0,>=1.20 in d:\anaconda\lib\site-packages (from seaborn) (1.26.4)

Requirement already satisfied: pandas>=1.2 in d:\anaconda\lib\site-packages (from seaborn) (2.2.2)

Requirement already satisfied: matplotlib!=3.6.1,>=3.4 in d:\anaconda\lib\site-packages (from seaborn) (3.8.4)

```

Requirement already satisfied: contourpy>=1.0.1 in d:\anaconda\lib\
site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (1.2.0)
Requirement already satisfied: cyclor>=0.10 in d:\anaconda\lib\site-
packages (from matplotlib!=3.6.1,>=3.4->seaborn) (0.11.0)
Requirement already satisfied: fonttools>=4.22.0 in d:\anaconda\lib\
site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (4.51.0)
Requirement already satisfied: kiwisolver>=1.3.1 in d:\anaconda\lib\
site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (1.4.4)
Requirement already satisfied: packaging>=20.0 in d:\anaconda\lib\
site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (23.2)
Requirement already satisfied: pillow>=8 in d:\anaconda\lib\site-
packages (from matplotlib!=3.6.1,>=3.4->seaborn) (10.3.0)
Requirement already satisfied: pyparsing>=2.3.1 in d:\anaconda\lib\
site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (3.0.9)
Requirement already satisfied: python-dateutil>=2.7 in d:\anaconda\
lib\site-packages (from matplotlib!=3.6.1,>=3.4->seaborn)
(2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in d:\anaconda\lib\site-
packages (from pandas>=1.2->seaborn) (2024.1)
Requirement already satisfied: tzdata>=2022.7 in d:\anaconda\lib\site-
packages (from pandas>=1.2->seaborn) (2023.3)
Requirement already satisfied: six>=1.5 in d:\anaconda\lib\site-
packages (from python-dateutil>=2.7->matplotlib!=3.6.1,>=3.4->seaborn)
(1.16.0)

```

#Quantity distribution by category

```

category_quantity = df.groupby('Category')
['Quantity'].sum().reset_index()
fig = px.pie(category_quantity, names='Category', values='Quantity',
title='Quantity Distribution by Category',
color_discrete_sequence=px.colors.qualitative.Pastel)
fig.show()

```

Quantity Distribution by Category



The pie chart shows the quantity percentage distribution amongst the categories.

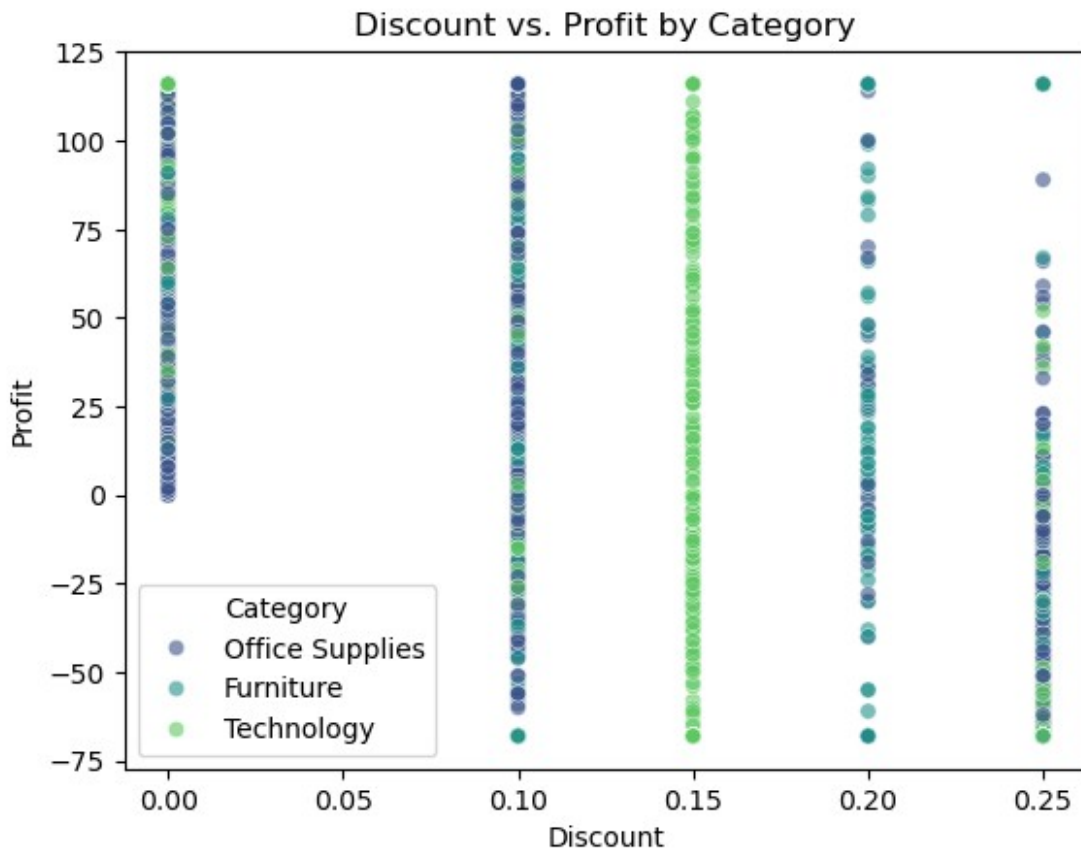
#Discount Impact on Profit

```

sns.scatterplot(data=df, x='Discount', y='Profit', hue='Category',

```

```
alpha=0.6, palette='viridis')
plt.title("Discount vs. Profit by Category")
plt.xlabel("Discount")
plt.ylabel("Profit")
plt.show()
```



Products with high discounts (> 25%) tend to have negative profits. Some categories, especially Office Supplies and Technology, show major losses when discounts are high.

```
#Sales and Profit by Region
region_sales_profit = df.groupby('Region')[['Sales',
'Profit']].sum().reset_index()

fig, ax = plt.subplots(1, 2, figsize=(14, 6))
sns.barplot(data=region_sales_profit, x='Region', y='Sales', ax=ax[0],
palette='Blues')
ax[0].set_title("Sales by Region")
ax[0].set_xlabel("Region")
ax[0].set_ylabel("Sales")

sns.barplot(data=region_sales_profit, x='Region', y='Profit',
ax=ax[1], palette='Greens')
ax[1].set_title("Profit by Region")
```

```
ax[1].set_xlabel("Region")
ax[1].set_ylabel("Profit")
```

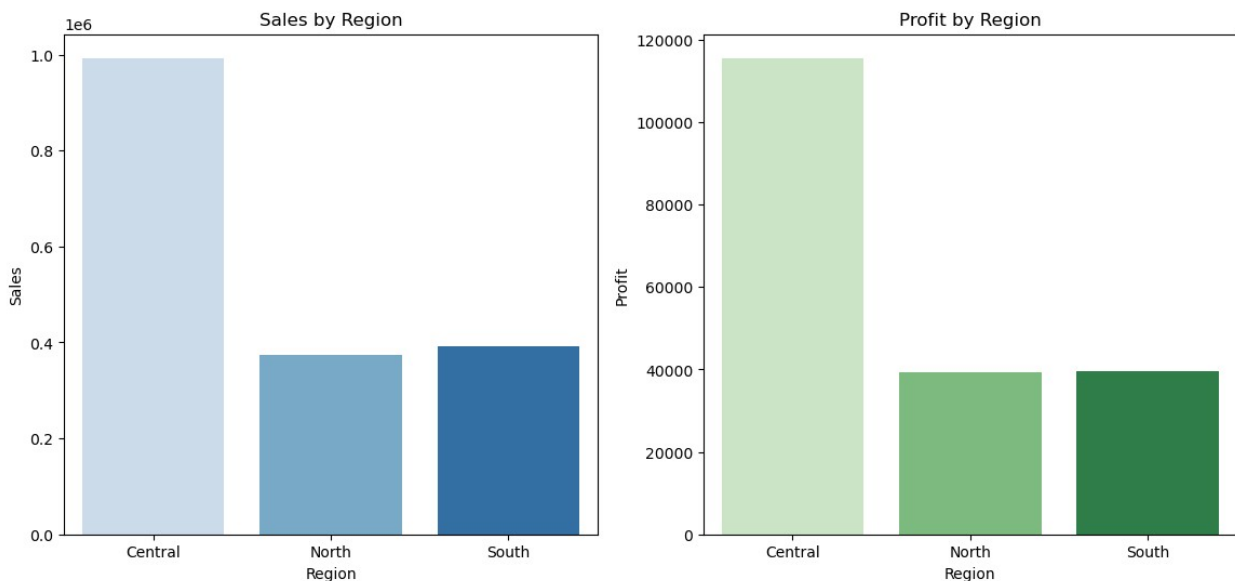
```
plt.show()
```

C:\Users\KIIT\AppData\Local\Temp\ipykernel_19476\828807372.py:4:
FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

C:\Users\KIIT\AppData\Local\Temp\ipykernel_19476\828807372.py:9:
FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.



The North region has the highest sales and profit. The Central region shows decent sales but low profit, meaning operational costs may be higher.

#Sales by Country distribution

```
category_quantity = df.groupby('Country')['Sales'].sum().reset_index()
fig = px.pie(category_quantity, names='Country', values='Sales',
```

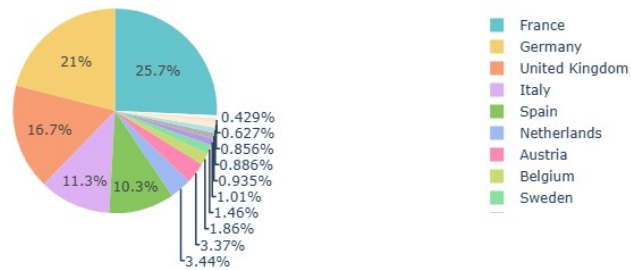


```

title='Total Sales by Country',
color_discrete_sequence=px.colors.qualitative.Pastel)
fig.show()

```

Total Sales by Country



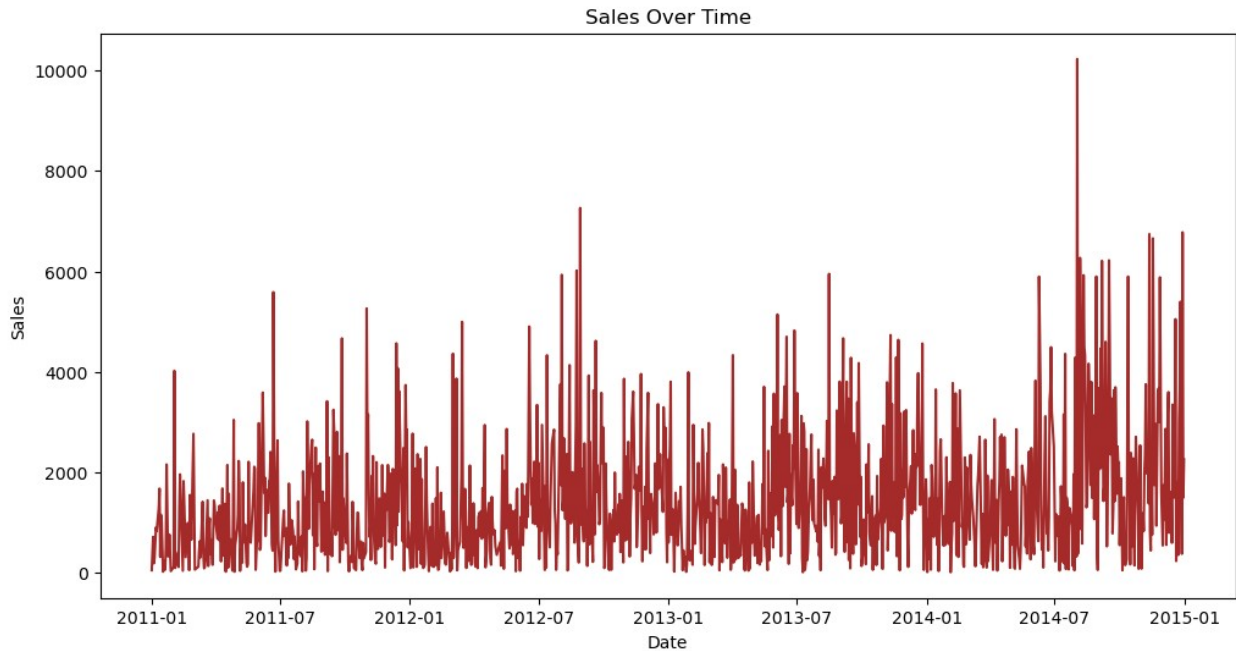
United Kingdom, Germany and France contribute the most to sales. Smaller markets like Netherlands and Ireland have minimal sales impact.

```

#Time Series Analysis
df['Order Date'] = pd.to_datetime(df['Order Date']) # Ensure it's in
datetime format
sales_over_time = df.groupby('Order Date')
['Sales'].sum().reset_index()

plt.figure(figsize=(12, 6))
sns.lineplot(data=sales_over_time, x='Order Date', y='Sales',
color='brown')
plt.title("Sales Over Time")
plt.xlabel("Date")
plt.ylabel("Sales")
plt.show()

```

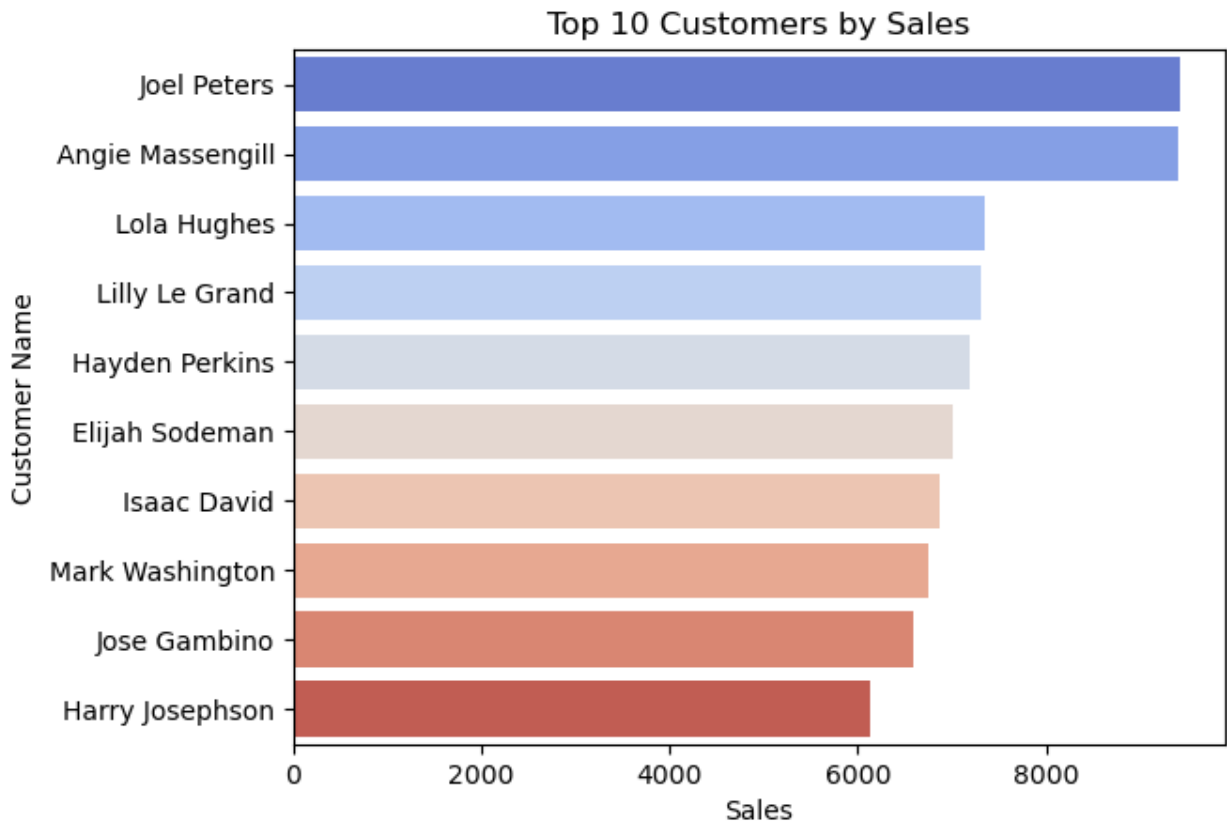


Sales show seasonal trends, with peaks in certain months.

```
#Top Customers by Sales
top_customers = df.groupby('Customer Name')
['Sales'].sum().nlargest(10).reset_index()
sns.barplot(data=top_customers, x='Sales', y='Customer Name',
palette='coolwarm')
plt.title("Top 10 Customers by Sales")
plt.xlabel("Sales")
plt.ylabel("Customer Name")
plt.show()
```

C:\Users\KIIT\AppData\Local\Temp\ipykernel_19476\1941503435.py:3:
FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

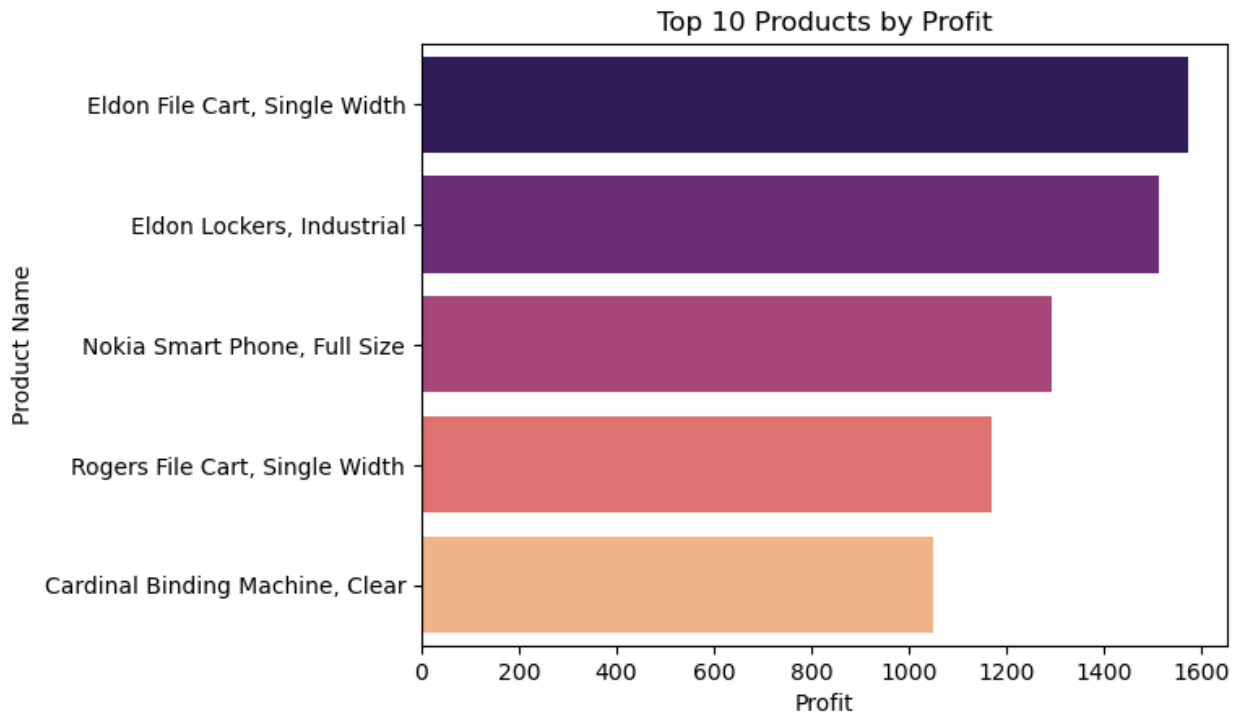


A few key customers contribute a large portion of total sales. These top customers likely make repeated high-value purchases.

```
#Top 5 products by Profit
top_products = df.groupby('Product Name')
['Profit'].sum().nlargest(5).reset_index()
sns.barplot(data=top_products, x='Profit', y='Product Name',
palette='magma')
plt.title("Top 10 Products by Profit")
plt.xlabel("Profit")
plt.ylabel("Product Name")
plt.show()
```

C:\Users\KIIT\AppData\Local\Temp\ipykernel_19476\315273585.py:3:
FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.



A few products generate the highest profits, indicating high-margin items. The most profitable products belong to Technology and Office Supplies categories.

Key Findings

Discounts negatively impact profit – A strong negative correlation (-0.49) indicates that excessive discounts reduce profitability. Optimizing discount strategies is crucial.

Some high-sales products generate minimal or negative profit, indicating a need for better pricing strategies.

The North region generates the highest sales and profit, while other regions need cost optimizations and better sales strategies.

There are clear end-of-year sales spikes, highlighting opportunities to align marketing efforts with peak demand.

A small group of customers accounts for a large portion of revenue, emphasizing the need for loyalty programs and personalized promotions.

These product categories have higher profitability, suggesting a focus on inventory management and targeted marketing for them.