## **E-commerce Data Insights**

- Which products and users generate the most revenue
- What times and patterns indicate peak performance

```
In [26]: # Import Required Libraries
   import pandas as pd
   import matplotlib.pyplot as plt
   import seaborn as sns
   import plotly.express as px
   %matplotlib inline
In [27]: # Load Dataset
   df = pd.read_excel("Superstore_USA.xlsx")
   df.head()
```

Out[27]:

•		Row ID	Order Priority	Discount	Unit Price	Shipping Cost	Customer ID	Customer Name	Ship Mode	Custome Segmer
	0	18606	Not Specified	0.01	2.88	0.50	2	Janice Fletcher	Regular Air	Corporat
	1	20847	High	0.01	2.84	0.93	3	Bonnie Potter	Express Air	Corporat
	2	23086	Not Specified	0.03	6.68	6.15	3	Bonnie Potter	Express Air	Corporat
	3	23087	Not Specified	0.01	5.68	3.60	3	Bonnie Potter	Regular Air	Corporat
	4	23088	Not Specified	0.00	205.99	2.50	3	Bonnie Potter	Express Air	Corporat

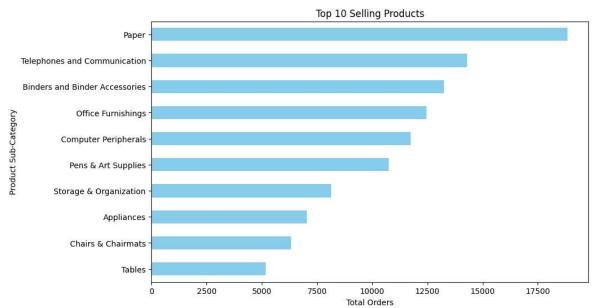
5 rows × 24 columns

In [42]: # to check whether a cell is having nulls or not
df.isnull().sum()

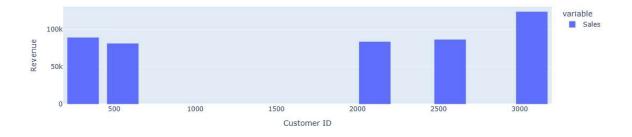
```
Out[42]: Row ID
                                   0
          Order Priority
                                   0
          Discount
                                   0
          Unit Price
                                   0
          Shipping Cost
                                   0
          Customer ID
                                   0
                                   0
          Customer Name
                                   0
          Ship Mode
          Customer Segment
                                   0
          Product Category
                                   0
          Product Sub-Category
                                   0
          Product Container
                                   0
          Product Name
                                   0
          Product Base Margin
                                   0
                                   0
          Region
          State or Province
                                   0
          City
                                   0
                                   0
          Postal Code
          Order Date
                                   0
          Ship Date
                                   0
          Profit
                                   0
          Quantity ordered new
                                   0
          Sales
                                   0
          Order ID
                                   0
          dtype: int64
```

## In [41]: df['Product Base Margin'] = df['Product Base Margin'].fillna(df['Product Base Ma

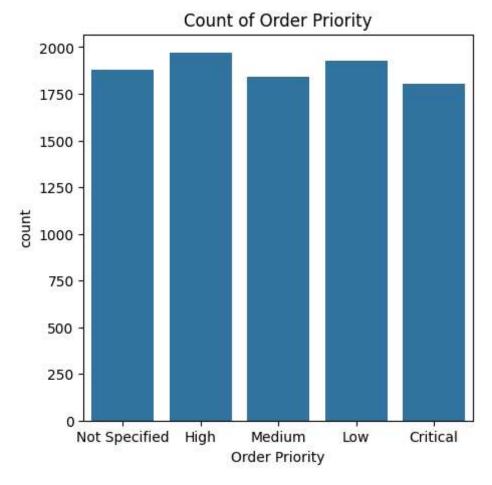
```
In [30]: # Top Selling Products
    top_products = df.groupby('Product Sub-Category')['Quantity ordered new'].sum().
    top_products.plot(kind='barh', title='Top 10 Selling Products', figsize=(10,6),
    plt.xlabel('Total Orders')
    plt.gca().invert_yaxis()
    plt.show()
```



```
In [31]: # Top Users by Revenue
top_users = df.groupby('Customer ID')['Sales'].sum().sort_values(ascending=False
fig = px.bar(top_users, title="Top 5 Revenue-Generating Users", labels={'value':
fig.show()
```



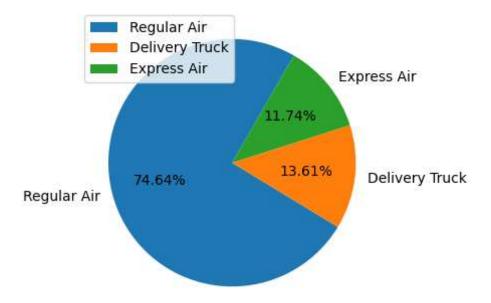
```
In [43]: # Order Priority
df['Order Priority'].unique()
Out[43]: array(['Not Specified', 'High', 'Medium', 'Low', 'Critical'], dtype=object)
In [33]: df['Order Priority'] = df["Order Priority"].replace("Critical ","Critical")
In [38]: plt.figure(figsize=(5,5))
    sns.countplot(x="Order Priority",data=df)
    plt.title("Count of Order Priority")
    plt.show()
```



```
In [44]: # Ship mode
In [39]: df['Ship Mode'].value_counts()
```

```
Delivery Truck 1283
Express Air 1107
Name: count, dtype: int64

In [40]: x = df['Ship Mode'].value_counts().index
y = df['Ship Mode'].value_counts().values
plt.figure(figsize=(5,4))
plt.pie(y,labels=x,startangle = 60,autopct="%0.2f%%")
plt.legend(loc=2)
plt.show()
```



## **Outcome**

Out[39]: Ship Mode

Regular Air

7036

• These insights help optimize pricing, marketing, and inventory strategy.

```
In [ ]:
```