

Data Collection and Preprocessing Phase

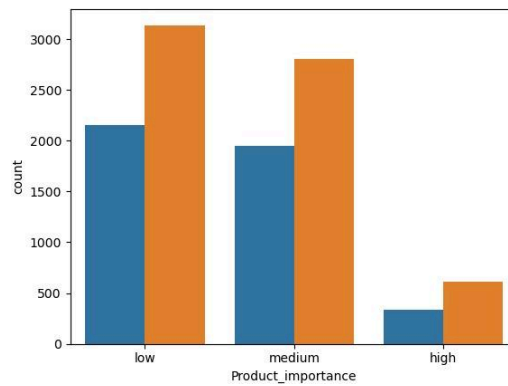
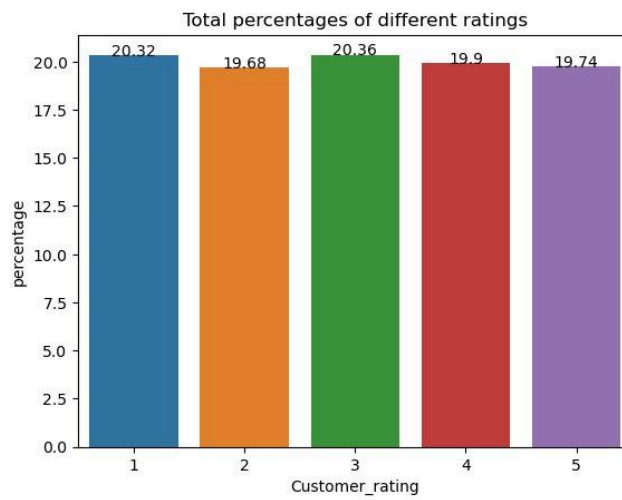
Date	11 JULY 2024
Team ID	SWTID1720116037
Project Title	Ecommerce Shipping Prediction Using Machine Learning
Maximum Marks	6 Marks

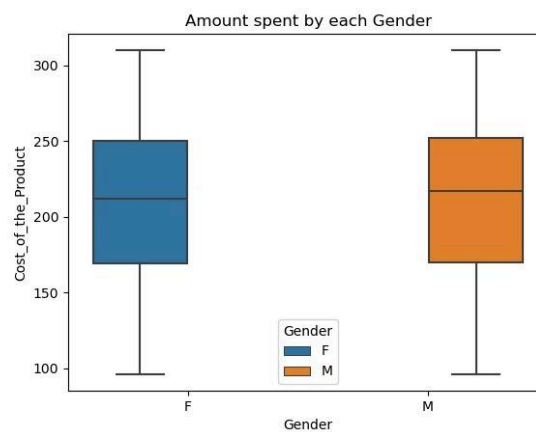
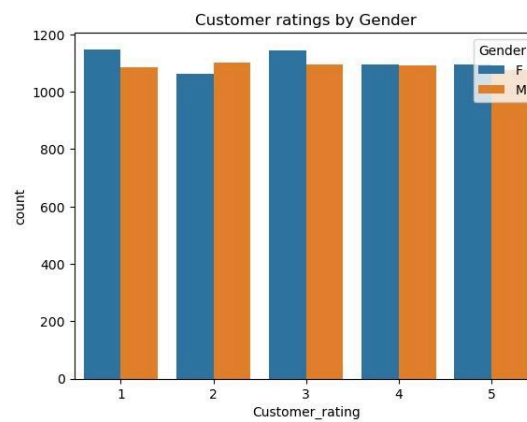
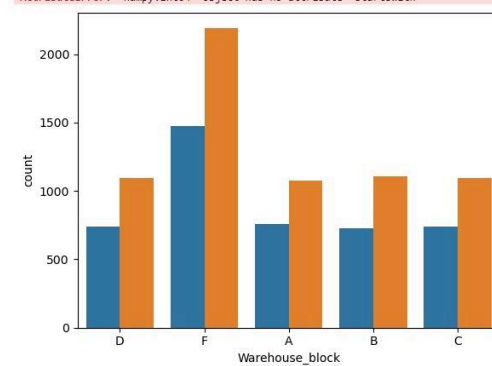
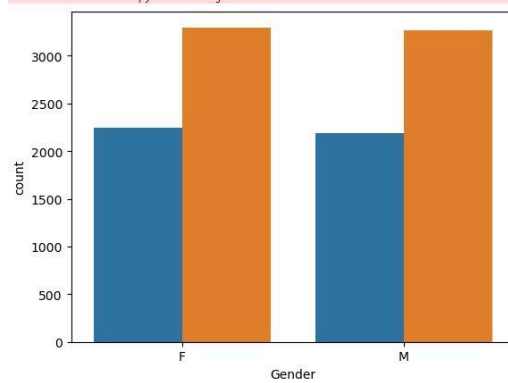
Data Exploration and Preprocessing Template

Identifies data sources, assesses quality issues like missing values and duplicates, and implements resolution plans to ensure accurate and reliable analysis.

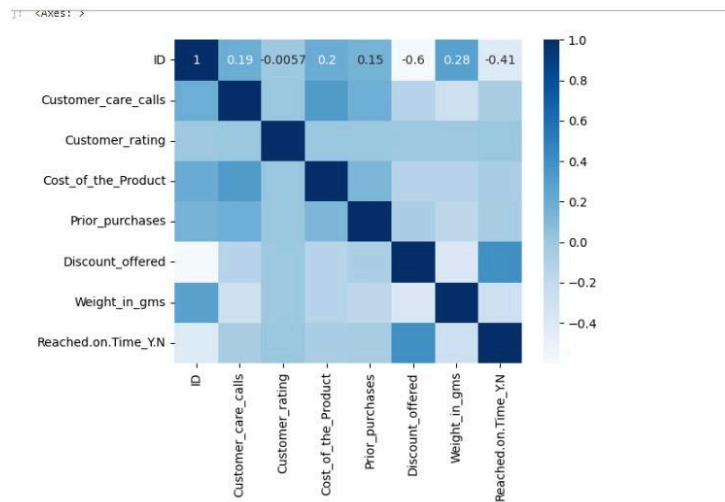
Section	Description																																																																																	
Data Overview	<table><thead><tr><th></th><th>ID</th><th>Customer_care_calls</th><th>Customer_rating</th><th>Cost_of_the_Product</th><th>Prior_purchases</th><th>Discount_offered</th><th>Weight_in_gms</th><th>Reached.on.Time_Y/N</th></tr></thead><tbody><tr><td>count</td><td>10999.000000</td><td>10999.000000</td><td>10999.000000</td><td>10999.000000</td><td>10999.000000</td><td>10999.000000</td><td>10999.000000</td><td>10999.000000</td></tr><tr><td>mean</td><td>5500.000000</td><td>4.054459</td><td>2.990545</td><td>210.196836</td><td>3.567597</td><td>13.373216</td><td>3634.016729</td><td>0.596691</td></tr><tr><td>std</td><td>3175.28214</td><td>1.141490</td><td>1.413603</td><td>48.063272</td><td>1.522860</td><td>16.205527</td><td>1635.377251</td><td>0.490584</td></tr><tr><td>min</td><td>1.000000</td><td>2.000000</td><td>1.000000</td><td>96.000000</td><td>2.000000</td><td>1.000000</td><td>1001.000000</td><td>0.000000</td></tr><tr><td>25%</td><td>2750.500000</td><td>3.000000</td><td>2.000000</td><td>169.000000</td><td>3.000000</td><td>4.000000</td><td>1839.500000</td><td>0.000000</td></tr><tr><td>50%</td><td>5500.000000</td><td>4.000000</td><td>3.000000</td><td>214.000000</td><td>3.000000</td><td>7.000000</td><td>4149.000000</td><td>1.000000</td></tr><tr><td>75%</td><td>6249.500000</td><td>5.000000</td><td>4.000000</td><td>251.000000</td><td>4.000000</td><td>10.000000</td><td>5050.000000</td><td>1.000000</td></tr><tr><td>max</td><td>10999.000000</td><td>7.000000</td><td>5.000000</td><td>310.000000</td><td>10.000000</td><td>65.000000</td><td>7846.000000</td><td>1.000000</td></tr></tbody></table>		ID	Customer_care_calls	Customer_rating	Cost_of_the_Product	Prior_purchases	Discount_offered	Weight_in_gms	Reached.on.Time_Y/N	count	10999.000000	10999.000000	10999.000000	10999.000000	10999.000000	10999.000000	10999.000000	10999.000000	mean	5500.000000	4.054459	2.990545	210.196836	3.567597	13.373216	3634.016729	0.596691	std	3175.28214	1.141490	1.413603	48.063272	1.522860	16.205527	1635.377251	0.490584	min	1.000000	2.000000	1.000000	96.000000	2.000000	1.000000	1001.000000	0.000000	25%	2750.500000	3.000000	2.000000	169.000000	3.000000	4.000000	1839.500000	0.000000	50%	5500.000000	4.000000	3.000000	214.000000	3.000000	7.000000	4149.000000	1.000000	75%	6249.500000	5.000000	4.000000	251.000000	4.000000	10.000000	5050.000000	1.000000	max	10999.000000	7.000000	5.000000	310.000000	10.000000	65.000000	7846.000000	1.000000
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Univariate Analysis	<div><div>Total number of delayed deliveries vs Shipment mode</div><table><thead><tr><th>Mode_of_Shipment</th><th>Percentage</th></tr></thead><tbody><tr><td>Flight</td><td>15.96</td></tr><tr><td>Road</td><td>16.34</td></tr><tr><td>Ship</td><td>67.7</td></tr></tbody></table></div>	Mode_of_Shipment	Percentage	Flight	15.96	Road	16.34	Ship	67.7																																																																									
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Bivariate Analysis

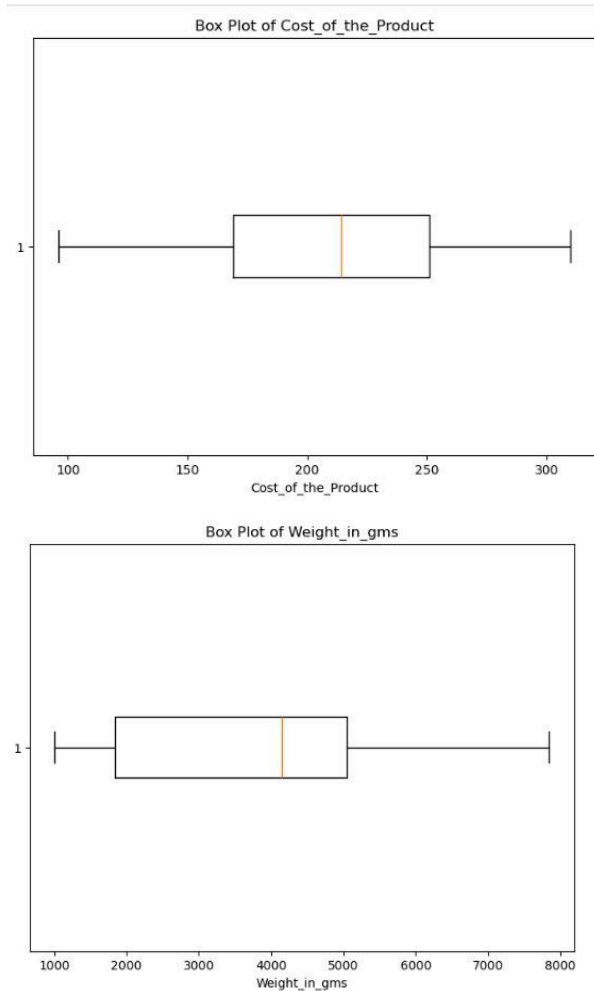




Multivariate Analysis



Outliers and Anomalies



Data Preprocessing Code Screenshots

Loading Data	<pre>import pandas as pd dataset = pd.read_csv('train.csv') dataset</pre>
Handling Missing Data	<pre>dataset.isnull().sum()</pre>
Data Transformation	<pre># Encode categorical variables le = LabelEncoder() dataset['Warehouse_block'] = le.fit_transform(dataset['Warehouse_block']) dataset['Mode_of_Shipment'] = le.fit_transform(dataset['Mode_of_Shipment']) dataset['Product_importance'] = le.fit_transform(dataset['Product_importance']) dataset['Gender'] = le.fit_transform(dataset['Gender']) # Scale/normalize features scaler = StandardScaler() columns_to_scale = ['Customer_care_calls', 'Customer_rating', 'Cost_of_the_Product', 'Prior_purchases', 'Discount_offered', 'Weight_in_gms'] dataset[columns_to_scale] = scaler.fit_transform(dataset[columns_to_scale])</pre>
Feature Engineering	<pre>import pandas as pd # create a sample dataframe data = {'priority': ['low', 'medium', 'high', 'low', 'medium', 'high']} dataset = pd.DataFrame(data) # create a new column with the mapped values dataset['priority_code'] = dataset['priority'].map({'low': 0, 'medium': 1, 'high': 2}) print(dataset)</pre>
Save Processed Data	<pre>dataset.to_csv('my_dataset.csv', index=False)</pre>