

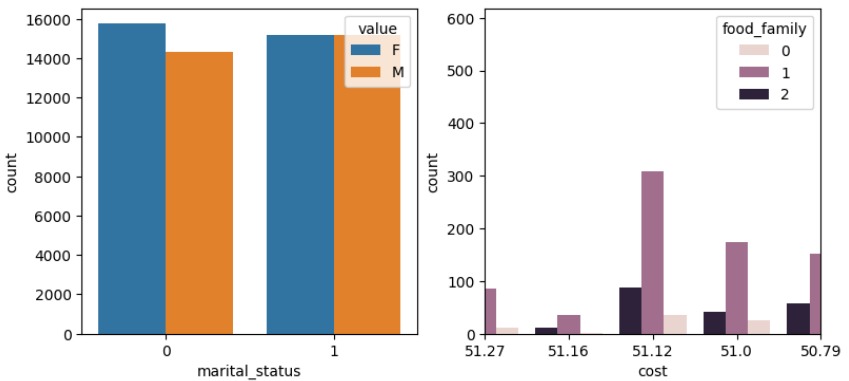
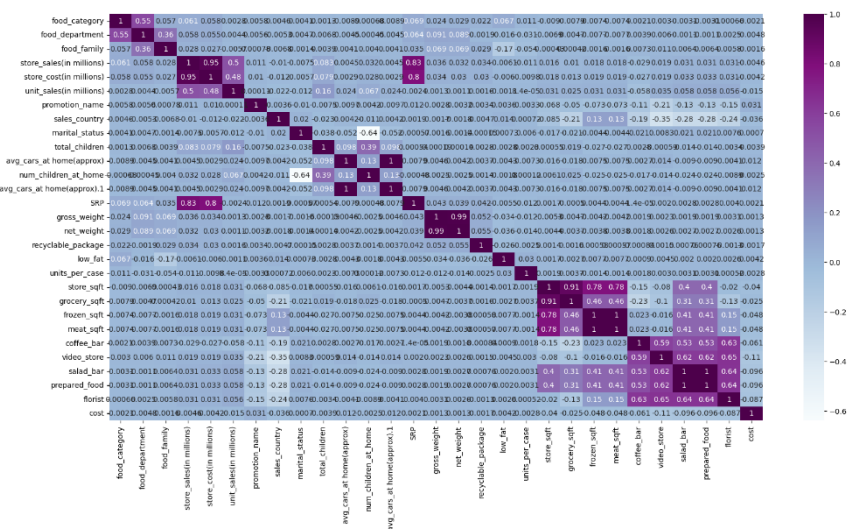
Data Collection and Preprocessing Phase

Date	4 th July 2024
Team ID	739804
Project Title	Cost Prediction of Acquiring a Customer
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	Dimensions: (60428, 40)																																																																																										
Univariate Analysis	<table><thead><tr><th></th><th>store_sales(in millions)</th><th>store_cost(in millions)</th><th>unit_sales(in millions)</th><th>total_children</th><th>avg_cars_at home (approx)</th><th>num_children_at_home</th><th>avg_cars_at home (approx).1</th><th>SRP</th><th>gross_weight</th></tr></thead><tbody><tr><td>count</td><td>60428.000000</td><td>60428.000000</td><td>60428.000000</td><td>60428.000000</td><td>60428.000000</td><td>60428.000000</td><td>60428.000000</td><td>60428.000000</td><td>60428.000000</td></tr><tr><td>mean</td><td>6.541031</td><td>2.619460</td><td>3.093169</td><td>2.533875</td><td>2.200271</td><td>0.829351</td><td>2.200271</td><td>2.115258</td><td>13.800000</td></tr><tr><td>std</td><td>3.463047</td><td>1.453009</td><td>0.827677</td><td>1.490165</td><td>1.109644</td><td>1.303424</td><td>1.109644</td><td>0.932829</td><td>4.622000</td></tr><tr><td>min</td><td>0.510000</td><td>0.163200</td><td>1.000000</td><td>0.000000</td><td>0.000000</td><td>0.000000</td><td>0.000000</td><td>0.500000</td><td>6.000000</td></tr><tr><td>25%</td><td>3.810000</td><td>1.500000</td><td>3.000000</td><td>1.000000</td><td>1.000000</td><td>0.000000</td><td>1.000000</td><td>1.410000</td><td>9.700000</td></tr><tr><td>50%</td><td>5.940000</td><td>2.385600</td><td>3.000000</td><td>3.000000</td><td>2.000000</td><td>0.000000</td><td>2.000000</td><td>2.130000</td><td>13.600000</td></tr><tr><td>75%</td><td>8.670000</td><td>3.484025</td><td>4.000000</td><td>4.000000</td><td>3.000000</td><td>1.000000</td><td>3.000000</td><td>2.790000</td><td>17.700000</td></tr><tr><td>max</td><td>22.920000</td><td>9.726500</td><td>6.000000</td><td>5.000000</td><td>4.000000</td><td>5.000000</td><td>4.000000</td><td>3.980000</td><td>21.900000</td></tr></tbody></table> <p>8 rows x 23 columns</p>		store_sales(in millions)	store_cost(in millions)	unit_sales(in millions)	total_children	avg_cars_at home (approx)	num_children_at_home	avg_cars_at home (approx).1	SRP	gross_weight	count	60428.000000	60428.000000	60428.000000	60428.000000	60428.000000	60428.000000	60428.000000	60428.000000	60428.000000	mean	6.541031	2.619460	3.093169	2.533875	2.200271	0.829351	2.200271	2.115258	13.800000	std	3.463047	1.453009	0.827677	1.490165	1.109644	1.303424	1.109644	0.932829	4.622000	min	0.510000	0.163200	1.000000	0.000000	0.000000	0.000000	0.000000	0.500000	6.000000	25%	3.810000	1.500000	3.000000	1.000000	1.000000	0.000000	1.000000	1.410000	9.700000	50%	5.940000	2.385600	3.000000	3.000000	2.000000	0.000000	2.000000	2.130000	13.600000	75%	8.670000	3.484025	4.000000	4.000000	3.000000	1.000000	3.000000	2.790000	17.700000	max	22.920000	9.726500	6.000000	5.000000	4.000000	5.000000	4.000000	3.980000	21.900000
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Multivariate Analysis	
Outliers and Anomalies	-
Data Preprocessing Code Screenshots	
Loading Data	<pre>df = pd.read_csv('/content/media prediction and its cost.csv')</pre>
Handling Missing Data	<pre>df.isnull().any()</pre>
Data Transformation	<pre>from sklearn.preprocessing import LabelEncoder def encoder(data, variable): # Corrected parameter name to 'variable' lb = LabelEncoder()</pre>

	<pre>df[variable] = lb.fit_transform(df[variable]) return lb</pre>
Feature Engineering	<pre>food_category_le = encoder(df,'food_category') brand_name_le = encoder(df,'brand_name') food_department_le = encoder(df,'food_department') food_family_le = encoder(df,'food_family') promotion_name_le = encoder(df,'promotion_name') store_city_le = encoder(df,'store_city') #unit_per_case_le = encoder(df,'unit_per_case') net_weight_le = encoder(df,'net_weight') sales_le = encoder(df,'sales_country') marital_le = encoder(df,'marital_status')</pre>
Save Processed Data	<pre>import pickle pickle.dump(rf,open('customers.pkl','wb')) pickle.dump(food_category_le,open('food_category_le.pkl','wb')) pickle.dump(brand_name_le,open('brand_name_le.pkl','wb')) pickle.dump(promotion_name_le,open('promotion_name_le.pkl','wb')) pickle.dump(store_city_le,open('store_city_le.pkl','wb'))</pre>