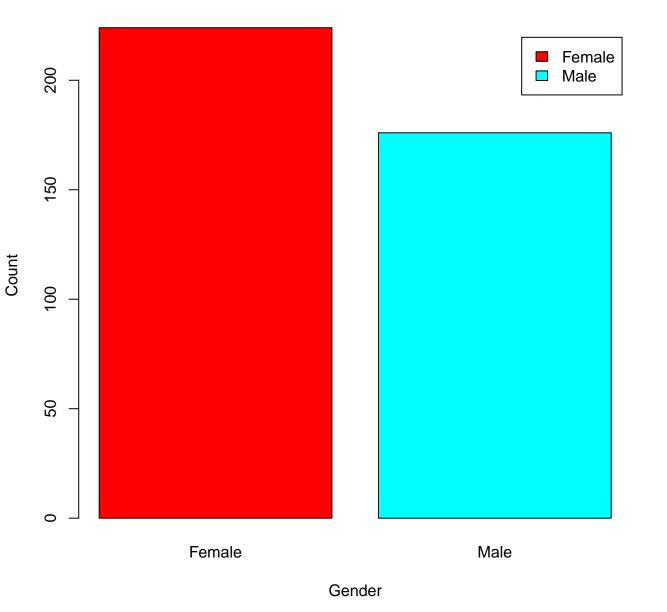
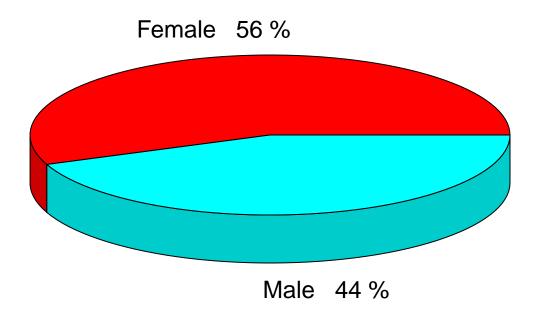
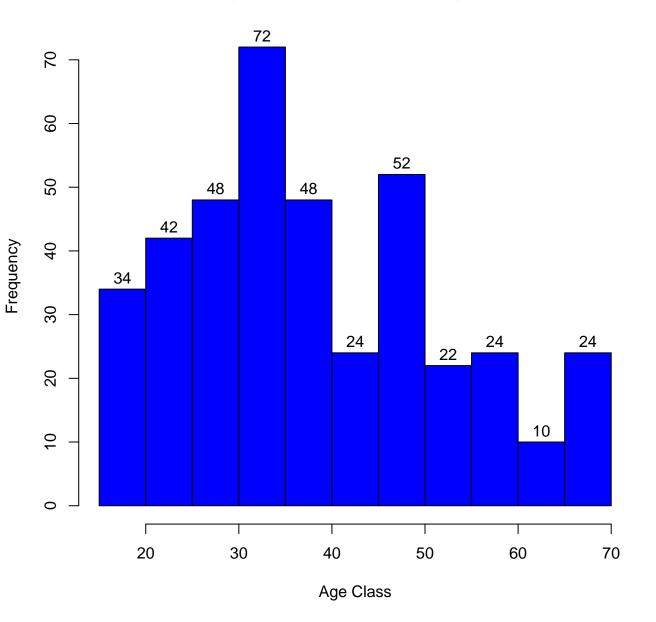
Using BarPlot to display Gender Comparision



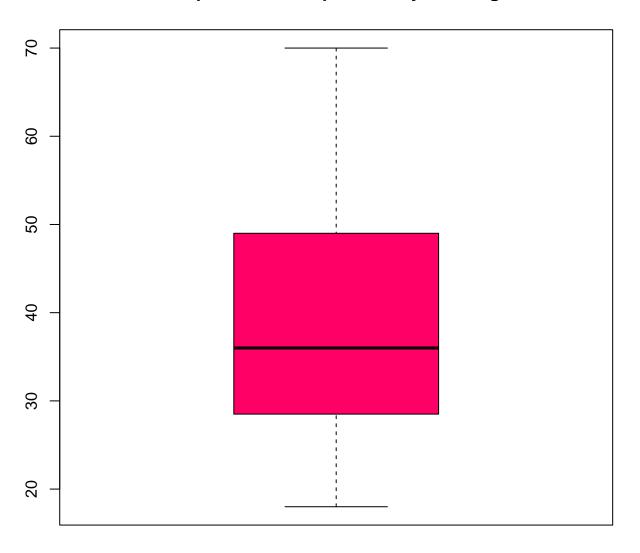
Pie Chart Depicting Ratio of Female and Male



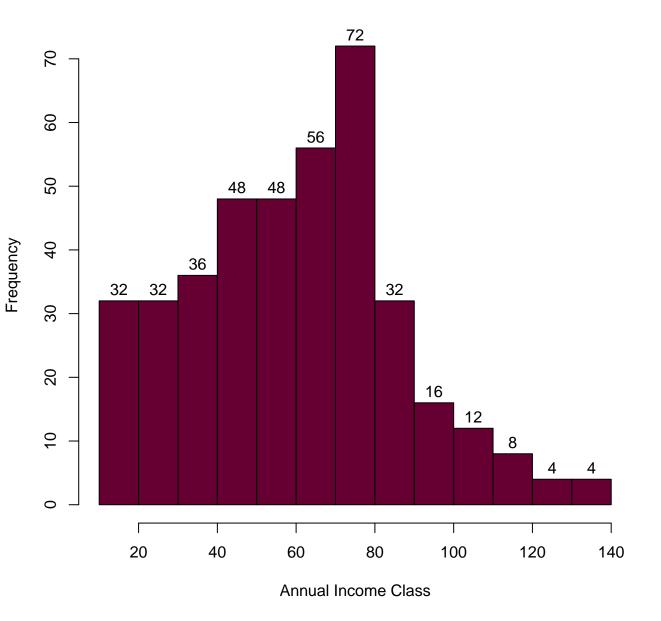
Histogram to Show Count of Age Class



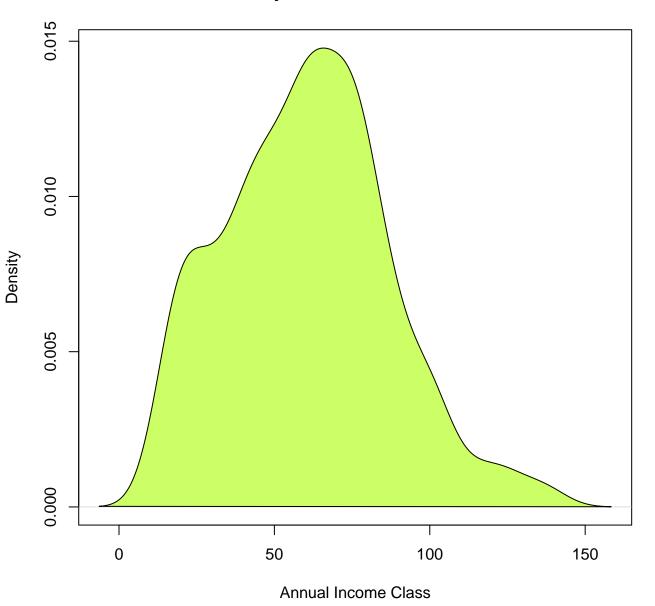
Boxplot for Descriptive Analysis of Age



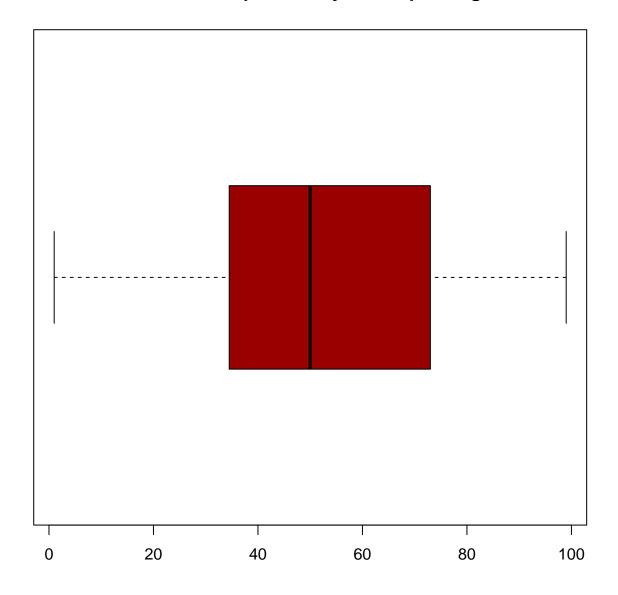
Histogram for Annual Income



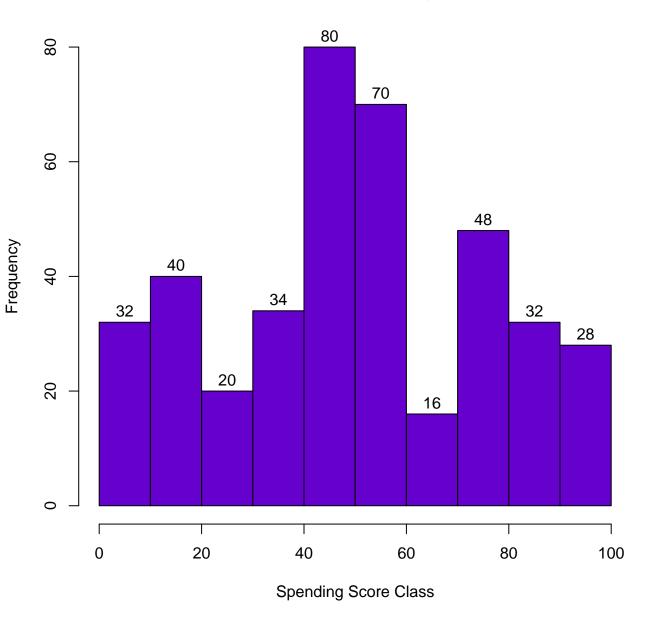
Density Plot for Annual Income

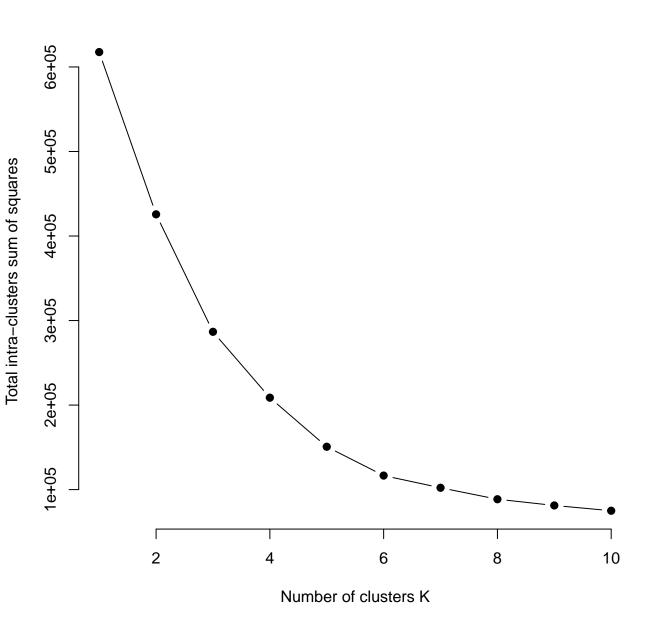


BoxPlot for Descriptive Analysis of Spending Score



HistoGram for Spending Score



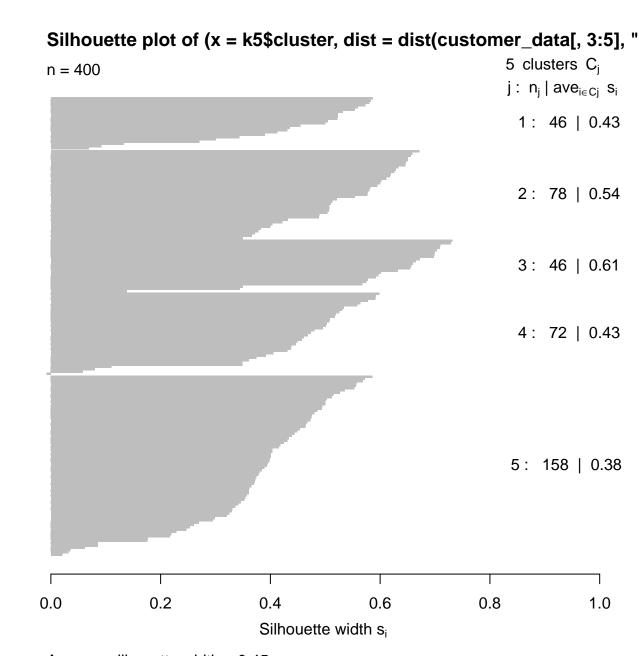


Silhouette plot of (x = k2\$cluster, dist = dist(customer_data[, 3:5], " 2 clusters C_i n = 400 $j:\ n_j \mid ave_{i \in Cj}\ s_i$ 1: 170 | 0.31 2: 230 | 0.29 0.0 0.2 0.4 0.6 8.0 1.0 Silhouette width si

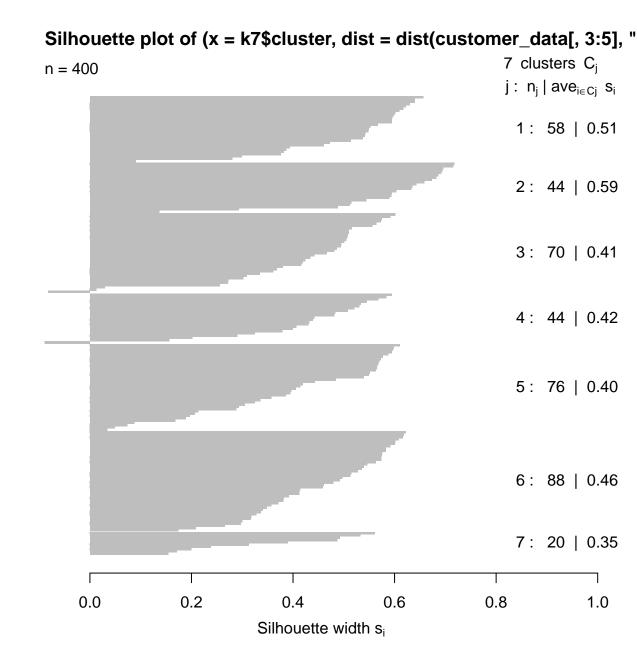
Silhouette plot of (x = k3\$cluster, dist = dist(customer_data[, 3:5], " 3 clusters C_i n = 400 $j: n_j \mid ave_{i \in C_j} s_i$ 1: 246 | 0.28 2: 76 | 0.50 3: 78 | 0.60 0.2 0.4 0.0 0.6 8.0 1.0

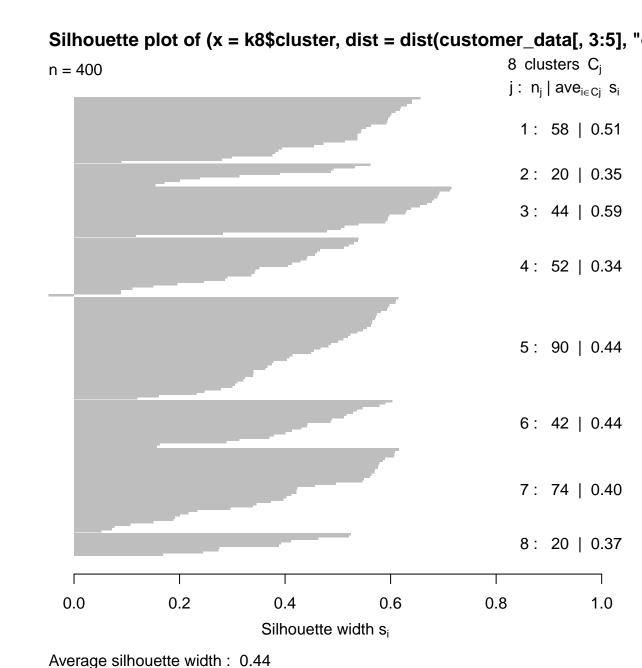
Silhouette width si

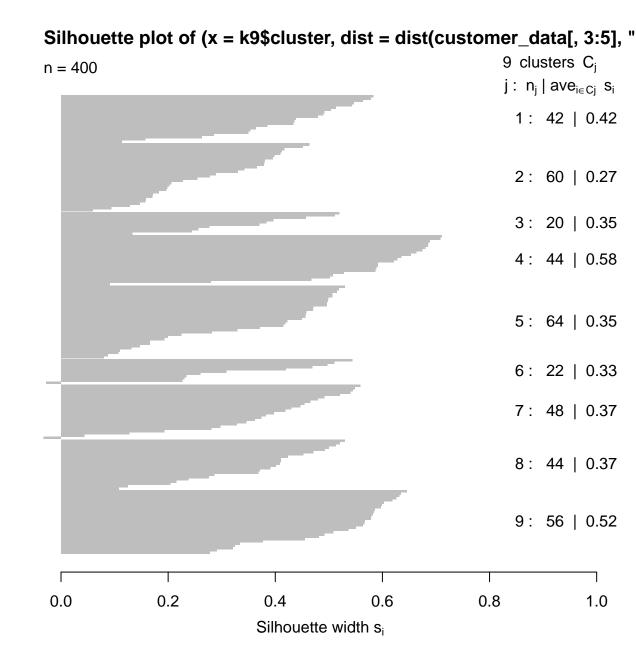
Silhouette plot of (x = k4\$cluster, dist = dist(customer_data[, 3:5], " 4 clusters C_i n = 400 $j: n_j \mid ave_{i \in C_j} s_i$ 1: 56 | 0.52 2: 78 | 0.59 3: 190 | 0.29 4: 76 | 0.44 0.0 0.4 0.2 0.6 8.0 1.0 Silhouette width si



Silhouette plot of (x = k6\$cluster, dist = dist(customer_data[, 3:5], " 6 clusters C_i n = 400 $j: n_j \mid ave_{i \in C_j} s_i$ 1: 78 | 0.51 2: 90 | 0.45 3: 42 | 0.44 4: 70 | 0.42 5: 44 | 0.59 6: 76 | 0.40 0.0 0.2 0.4 0.6 8.0 1.0 Silhouette width si

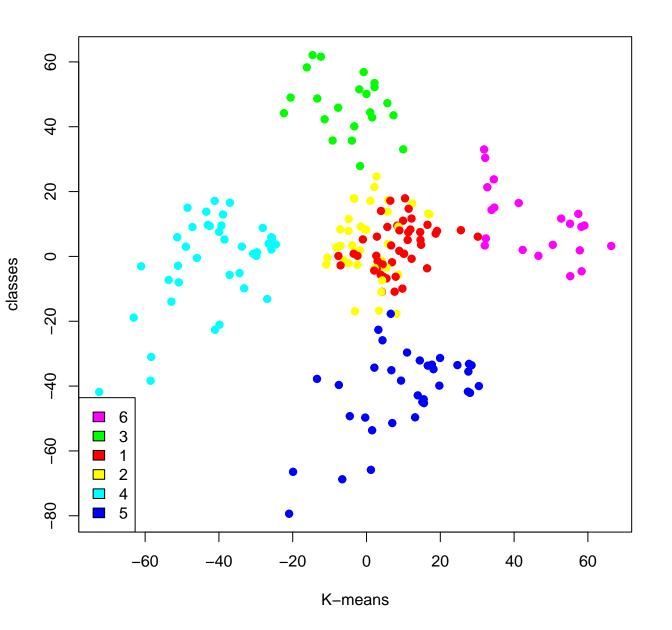


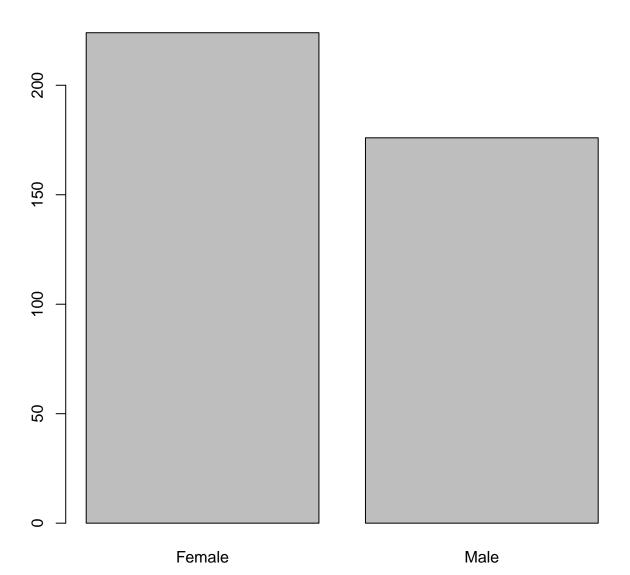




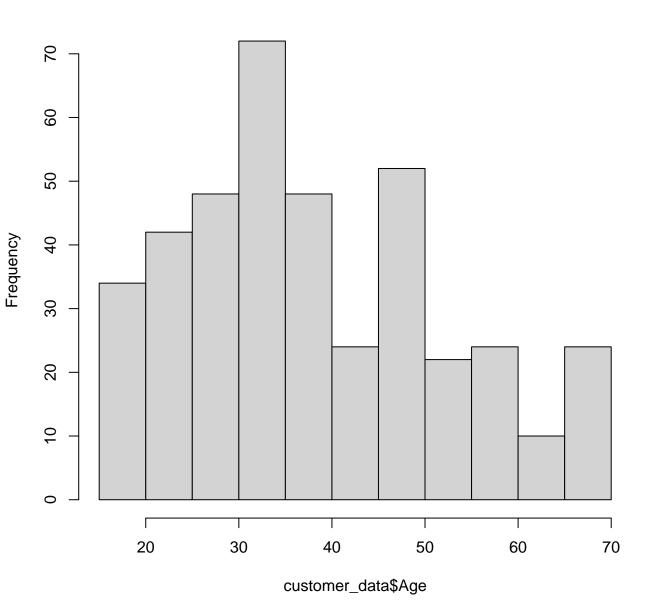
Silhouette plot of $(x = k10\cluster, dist = dist(customer_data[, 3:5],$ 10 clusters C_i n = 400 $j: n_j \mid ave_{i \in C_j} s_i$ 1: 56 | 0.51 2: 58 | 0.38 3: 26 | 0.31 4: 22 | 0.33 5: 54 | 0.32 6: 26 | 0.39 7: 44 | 0.57 8: 48 | 0.33 9: 44 | 0.40 10: 22 | 0.32 0.0 0.2 0.4 0.6 8.0 1.0

Silhouette width si

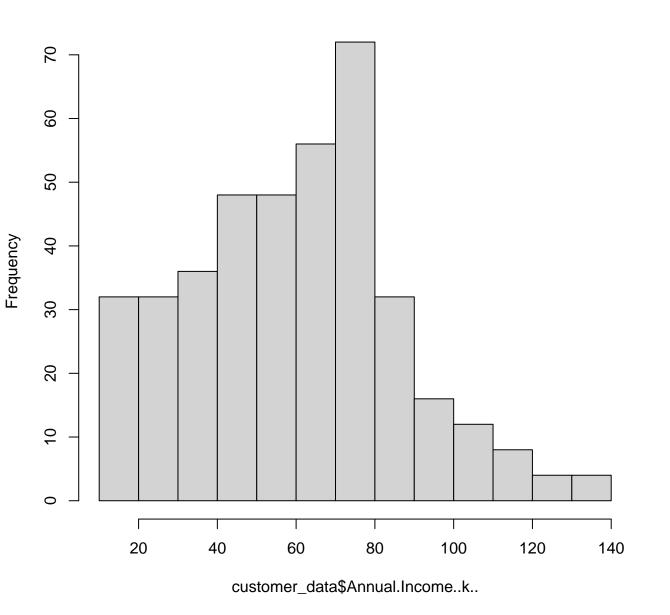




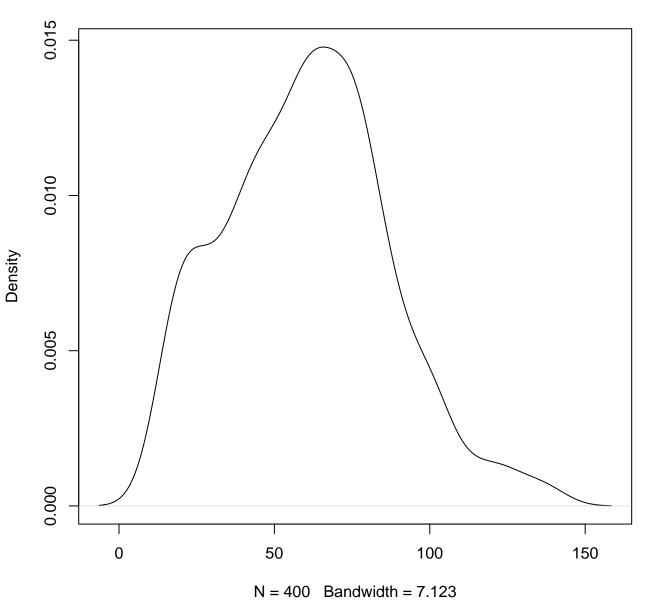
Histogram of customer_data\$Age

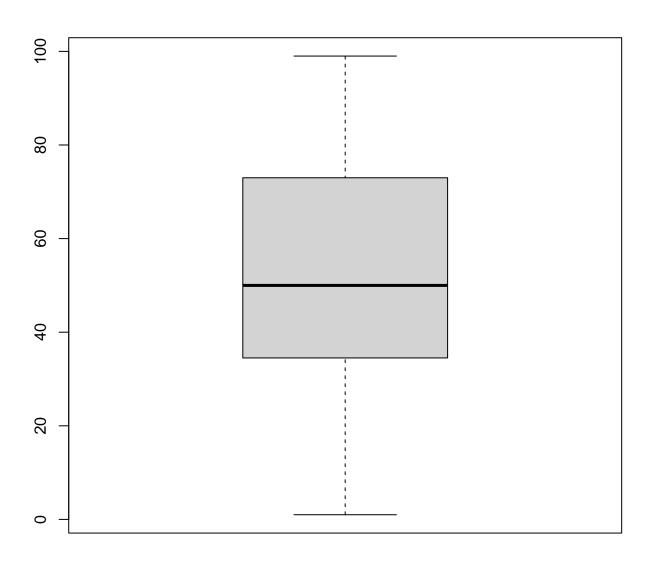


Histogram of customer_data\$Annual.Income..k..

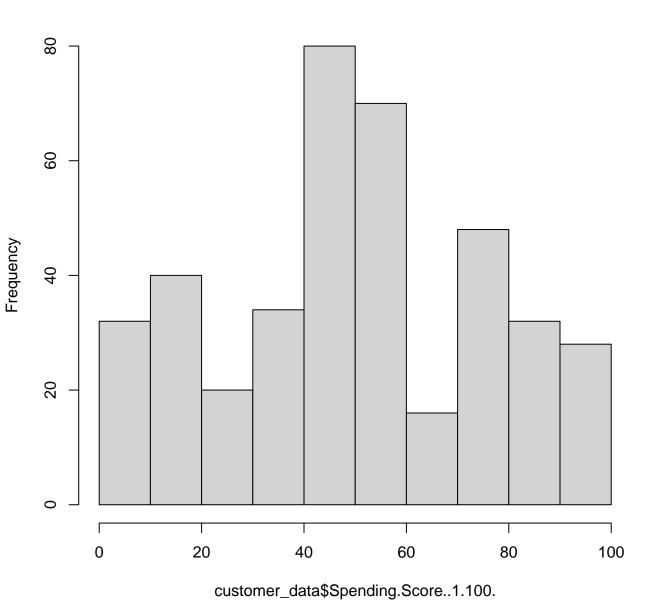


density(x = customer_data\$Annual.Income..k..)





Histogram of customer_data\$Spending.Score..1.100.



Elbow Method for Optimal Number of Clusters 1000 -Total Within-Cluster Sum of Squares 750 **-**500 -250 -5.0 7.5 10.0 2.5 Number of Clusters K