

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.decomposition import PCA
from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.metrics import silhouette_score
```

```
In [59]: mcdonalds = pd.read_csv("/kaggle/input/mcdonald-csv-file/mcdonalds.csv")
```

```
In [60]: print(f"Columns in dataframe: {mcdonalds.columns}")
print(f"Shape of dataframe: {mcdonalds.shape}")
print(mcdonalds.head(3))
```

```
Columns in dataframe: Index(['yummy', 'convenient', 'spicy', 'fattening', 'greasy',
                             'fast', 'cheap', 'tasty', 'expensive', 'healthy',
                             'disgusting', 'Like', 'Age', 'VisitFrequency', 'Gender'],
                             dtype='object')
```

```
Shape of dataframe: (1453, 15)
```

	yummy	convenient	spicy	fattening	greasy	fast	cheap	tasty	expensive	healthy
0	No	Yes	No	Yes	No	Yes	Yes	No	Yes	No
1	Yes	Yes	No	Yes	Yes	Yes	Yes	Yes	Yes	No
2	No	Yes	Yes	Yes	Yes	Yes	No	Yes	Yes	Yes

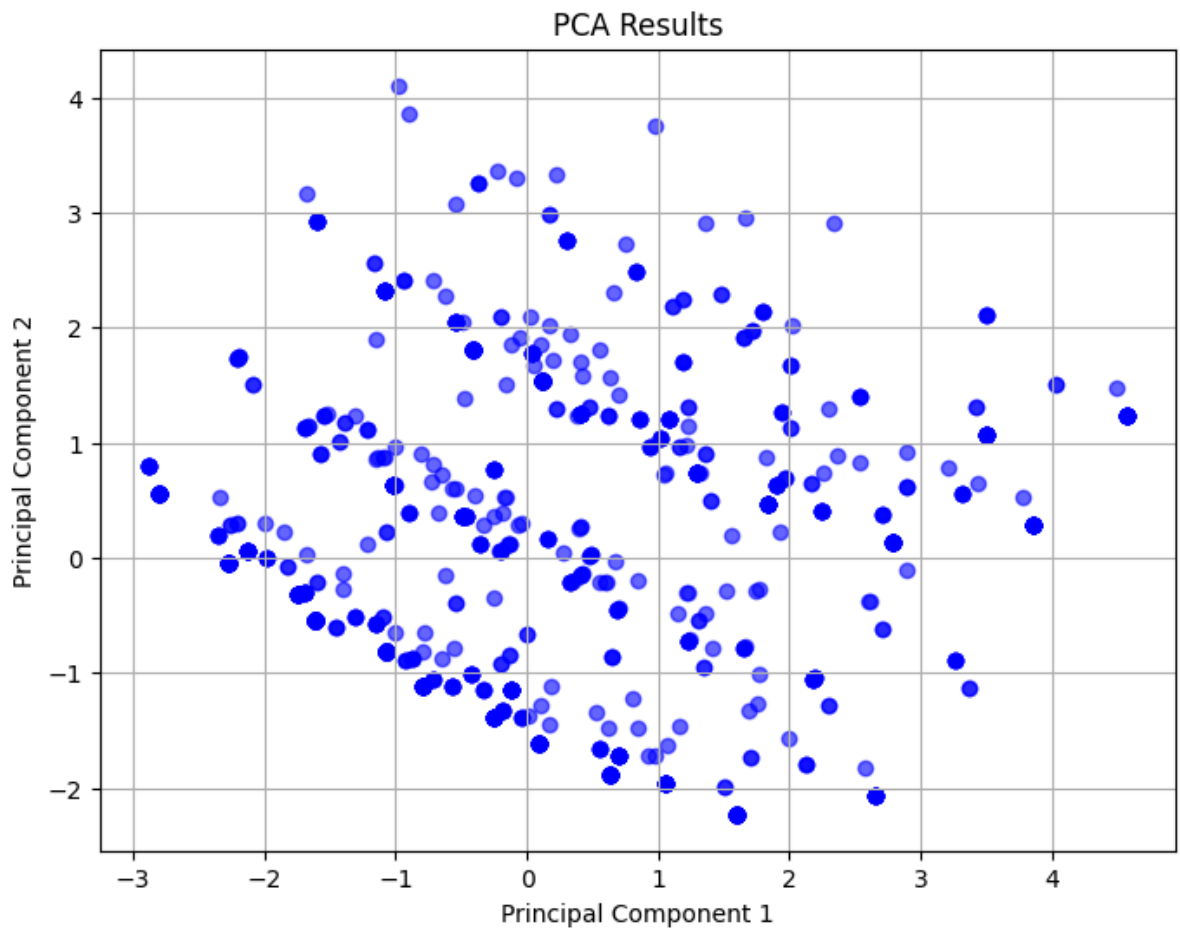
	disgusting	Like	Age	VisitFrequency	Gender
0	No	-3	61	Every three months	Female
1	No	+2	51	Every three months	Female
2	No	+1	62	Every three months	Female

```
In [61]: # Step 6: Profiling Segments
# Data Preprocessing
MD_x = mcdonalds.iloc[:, :11]
MD_x = (MD_x == "Yes").astype(int) # Convert categorical responses to binary
```

```
In [62]: # Normalize the data for PCA
scaler = StandardScaler()
MD_x_scaled = scaler.fit_transform(MD_x)
```

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In [63]: pca = PCA(n_components=2)
MD_x_pca = pca.fit_transform(MD_x_scaled)
```

```
In [64]: plt.figure(figsize=(8, 6))
plt.scatter(MD_x_pca[:, 0], MD_x_pca[:, 1], alpha=0.6, color='b')
plt.title("PCA Results")
plt.xlabel("Principal Component 1")
plt.ylabel("Principal Component 2")
plt.grid(True)
plt.show()
```

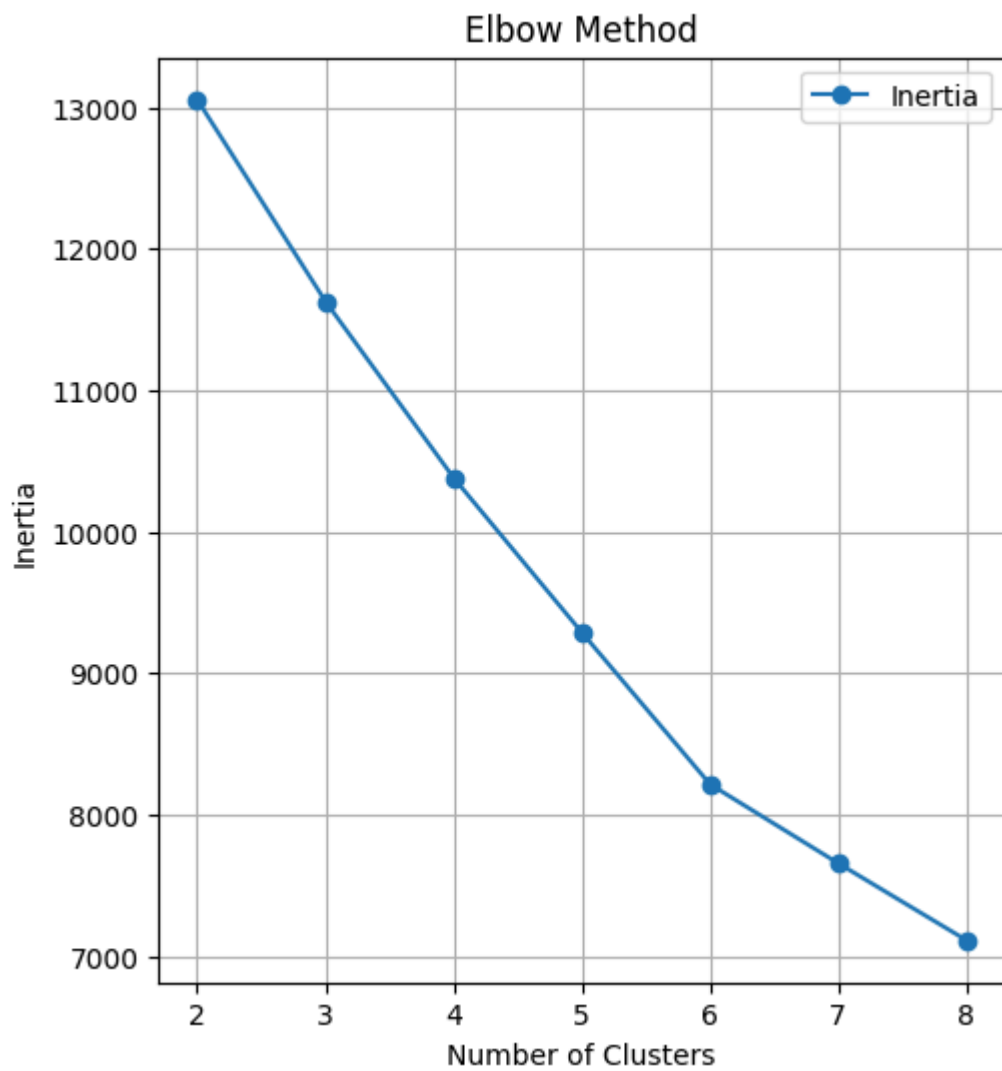


```
In [65]: # Perform K-means clustering
inertia = []
silhouette_scores = []
range_clusters = range(2, 9)
```

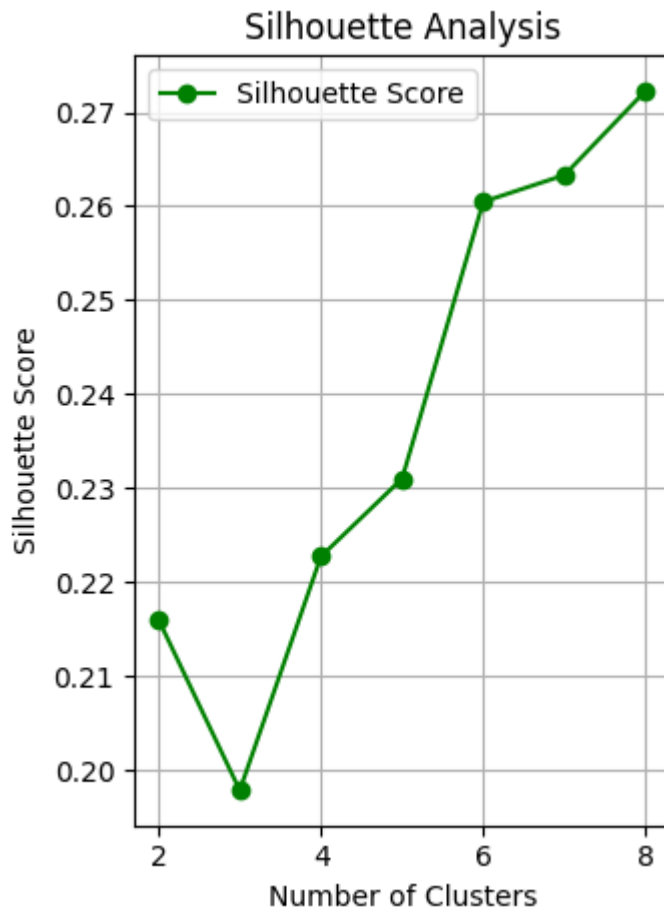
```
In [66]: for k in range_clusters:
kmeans = KMeans(n_clusters=k, random_state=1234, n_init=10)
kmeans.fit(MD_x_scaled)
inertia.append(kmeans.inertia_)
silhouette_scores.append(silhouette_score(MD_x_scaled, kmeans.labels_))
```

```
In [67]: # Plot inertia (Elbow Method) and Silhouette Scores
plt.figure(figsize=(12, 6))
plt.subplot(1, 2, 1)
plt.plot(range_clusters, inertia, marker='o', label='Inertia')
plt.title("Elbow Method")
plt.xlabel("Number of Clusters")
plt.ylabel("Inertia")
plt.grid(True)
plt.legend()
```

```
Out[67]: <matplotlib.legend.Legend at 0x7959e42bd4b0>
```



```
In [68]: plt.subplot(1, 2, 2)
plt.plot(range_clusters, silhouette_scores, marker='o', label='Silhouette Score', c
plt.title("Silhouette Analysis")
plt.xlabel("Number of Clusters")
plt.ylabel("Silhouette Score")
plt.grid(True)
plt.legend()
plt.tight_layout()
plt.show()
```



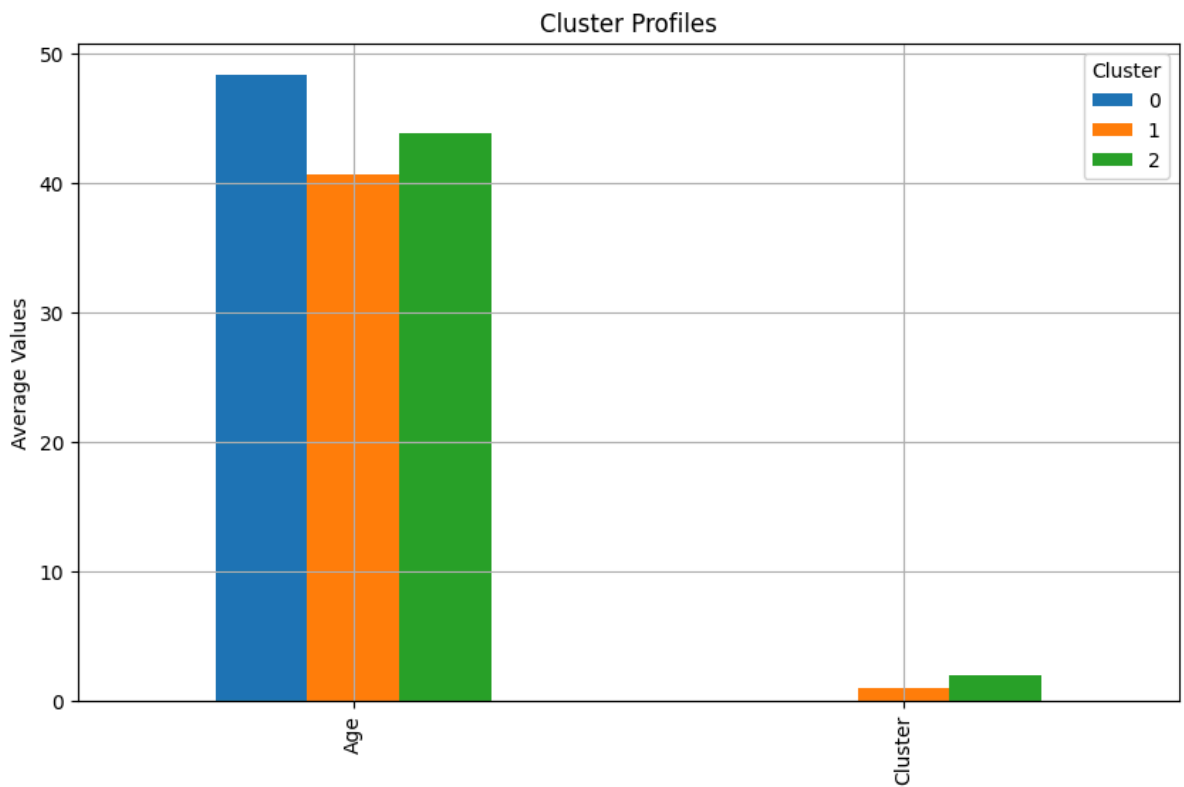
```
In [69]: optimal_k = 3
kmeans = KMeans(n_clusters=optimal_k, random_state=1234, n_init=10)
kmeans.fit(MD_x_scaled)
cluster_labels = kmeans.labels_
```

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In [70]: mcdonalds['Cluster'] = cluster_labels
```

```
In [71]: # Selecting the Target Segment
# Profiling clusters
numeric_columns = mcdonalds.select_dtypes(include=['number']).columns
cluster_profile = mcdonalds.groupby('Cluster')[numeric_columns].mean()
print("Cluster Profiles:")
print(cluster_profile)
```

```
Cluster Profiles:
           Age  Cluster
Cluster
0      48.394572      0.0
1      40.692529      1.0
2      43.880192      2.0
```

```
In [72]: cluster_profile.T.plot(kind='bar', figsize=(10, 6))
plt.title("Cluster Profiles")
plt.ylabel("Average Values")
plt.grid(True)
plt.show()
```



```
In [77]: # Customizing the Marketing Mix
for cluster in range(optimal_k):
    print(f"\nCluster {cluster}: Recommendations")

    cluster_data = cluster_profile.iloc[cluster]

    products_to_focus = cluster_data[cluster_data > 0.5].index
    less_popular_items = cluster_data[cluster_data <= 0.5].index

    print("- Products to focus on:", list(products_to_focus))
    print("- Less popular items:", list(less_popular_items))
```

Cluster 0: Recommendations

- Products to focus on: ['Age']
- Less popular items: ['Cluster']

Cluster 1: Recommendations

- Products to focus on: ['Age', 'Cluster']
- Less popular items: []

Cluster 2: Recommendations

- Products to focus on: ['Age', 'Cluster']
- Less popular items: []

```
In [78]: output_path = "/kaggle/working/mcdonalds_with_clusters.csv"
mcdonalds.to_csv(output_path, index=False)
print(f"Clustered dataset saved to: {output_path}")
```

Clustered dataset saved to: /kaggle/working/mcdonalds_with_clusters.csv