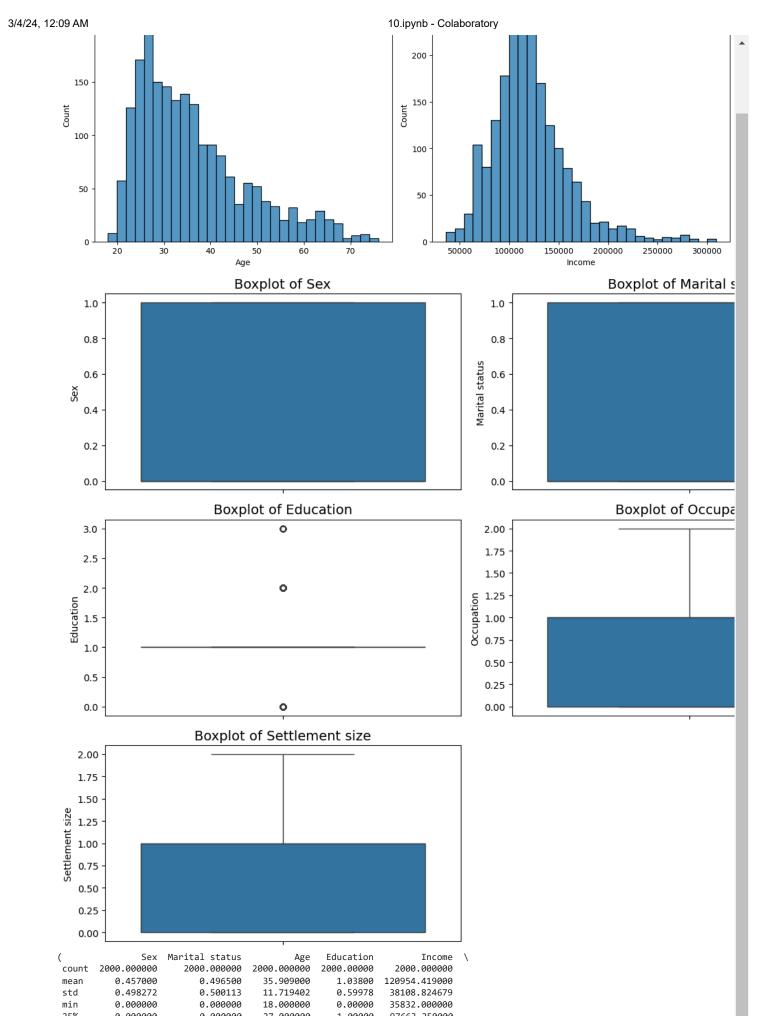
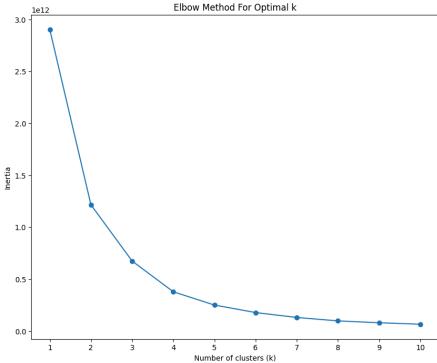
```
import pandas as pd
# Load the data excluding the 'ID' column
cust_data = pd.read_csv('segmentation data.csv', usecols=['Sex', 'Marital status', 'Age', 'Education', 'Income', 'Occupation', 'Settlem
cust_data.head()
df = cust_data
from sklearn.preprocessing import StandardScaler
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
import seaborn as sns
df = pd.read_csv('segmentation data.csv')
df_cluster = df.drop('ID', axis=1)
# Checking for missing values
missing_values = df_cluster.isnull().sum()
# Checking data types
data_types = df_cluster.dtypes
descriptive_stats = df_cluster.describe()
# Checking for any missing values in the dataset
missing_data = df_cluster.isnull().sum()
# Plotting histograms for the numerical variables
plt.figure(figsize=(12,10))
# Iterate over each numerical feature to create a histogram
for index, feature in enumerate(['Age', 'Income']):
    plt.subplot(2, 2, index+1)
    sns.histplot(df_cluster[feature], kde=False, bins=30)
    plt.title(f'Distribution of {feature}', size=14)
    plt.tight_layout()
plt.figure(figsize=(12,10))
for index, feature in enumerate(['Sex', 'Marital status', 'Education', 'Occupation', 'Settlement size']):
    plt.subplot(3, 2, index+1)
    sns.boxplot(y=df_cluster[feature])
    plt.title(f'Boxplot of {feature}', size=14)
    plt.tight_layout()
plt.show()
descriptive_stats, missing_data
```



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```

```
import pandas as pd
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
inertia = []
k_values = range(1, 11)
\# Loop over each k value, fit the KMeans model, and add the inertia to the list
for k in k_values:
    kmeans = KMeans(n_clusters=k, random_state=42)
    kmeans.fit(df)
    inertia.append(kmeans.inertia_)
# Plot the elbow curve
plt.figure(figsize=(10, 8))
plt.plot(k_values, inertia, marker='o')
plt.title('Elbow Method For Optimal k')
plt.xlabel('Number of clusters (k)')
plt.ylabel('Inertia')
plt.xticks(k_values)
plt.show()
```

```
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```



```
from sklearn.metrics import silhouette_score

# Fit K-Means with the chosen number of clusters
kmeans = KMeans(n_clusters=3, random_state=42)
kmeans.fit(df.drop('Cluster', axis=1))

# Assign the clusters to the dataframe
df['Cluster'] = kmeans.labels_

# Calculate silhouette score
silhouette_avg = silhouette_score(df.drop('Cluster', axis=1), kmeans.labels_)
print(f'Silhouette Score: {silhouette_avg:.2f}')

cluster_centers = pd.DataFrame(kmeans.cluster_centers_, columns=df.columns[:-1])
print(cluster_centers)
```

```
/usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change fr
  warnings.warn(
Silhouette Score: 0.51
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```

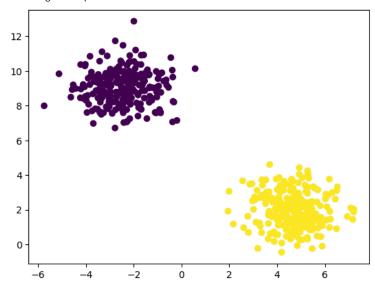
```
from sklearn.datasets import make_blobs
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt

# Generate synthetic data with 2 centers
features, _ = make_blobs(n_samples=500, centers=2, random_state=42)

# Run K-Means clustering
kmeans = KMeans(n_clusters=2, random_state=42)
kmeans.fit(features)

# Plot the clustered data
plt.scatter(features[:, 0], features[:, 1], c=kmeans.labels_, cmap='viridis')
plt.show()
```

/usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarninq warnings.warn(



```
import pandas as pd
from sklearn.cluster import KMeans
from sklearn.metrics import silhouette_score
import matplotlib.pyplot as plt
# Assuming df is your DataFrame after loading 'segmentation data.csv'
\# Fit K-Means and calculate silhouette scores for k{=}2 and k{=}3
for k in [2, 3]:
    kmeans = KMeans(n_clusters=k, random_state=42)
    kmeans.fit(df)
    silhouette_avg = silhouette_score(df, kmeans.labels_)
    print(f'Silhouette Score for k={k}: {silhouette_avg:.2f}')
\# Calculate inertia for a range of k values to plot the elbow graph
inertia = []
k_values = range(1, 11)
for k in k_values:
    kmeans = KMeans(n_clusters=k, random_state=42)
    kmeans.fit(df)
```

```
inertia.append(kmeans.inertia_)
```

```
# Plot the elbow curve
plt.figure(figsize=(10, 8))
plt.plot(k_values, inertia, marker='o')
plt.title('Elbow Method For Optimal k')
plt.xlabel('Number of clusters (k)')
plt.ylabel('Inertia')
plt.xticks(k_values)
plt.show()
```

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