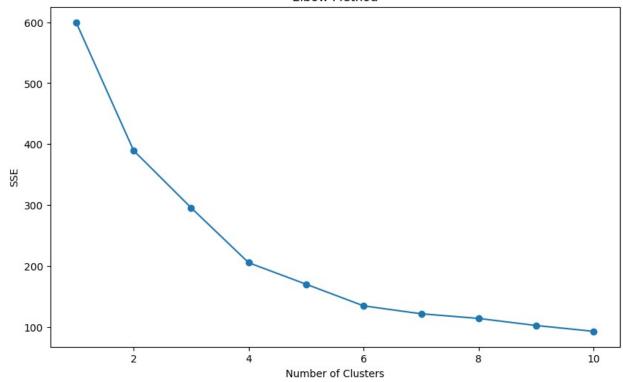
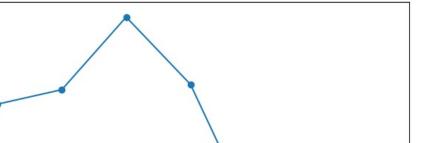
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import sklearn.preprocessing
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.cluster import KMeans
from sklearn.metrics import silhouette score
optimal k = 5
data = pd.read_csv('/content/Mall_Customers.csv')
numeric columns = data.select dtypes(include=[np.number]).columns
non numeric columnms = data.select dtypes(exclude=[np.number]).columns
data[numeric columns] =
data[numeric_columns].fillna(data[numeric_columns].mean())
label encoder = LabelEncoder()
for column in data.select dtypes(exclude=[np.number]).columns:
    data[column] =
label encoder.fit transform(data[column].astype(str))
features = ['Age', 'Annual Income (k$)', 'Spending Score (1-100)']
X = data[features]
scaler = StandardScaler()
X scaled = scaler.fit transform(X)
sse = []
for k in range(1, 11):
    kmeans = KMeans(n clusters=k, random state=42)
    kmeans.fit(X scaled)
    sse.append(kmeans.inertia )
plt.figure(figsize=(10, 6))
plt.plot(range(1, 11), sse ,marker='o')
plt.xlabel('Number of Clusters')
plt.vlabel('SSE')
plt.title('Elbow Method')
plt.show()
```

Elbow Method



```
silhouette_scores = []
for k in range(2, 11):
    kmeans = KMeans(n_clusters=optimal_k, random_state=42)
    kmeans.fit(X_scaled)
    silhouette_scores.append(silhouette_score(X_scaled,
kmeans.labels_))

plt.figure(figsize=(10, 6))
plt.plot(range(2, 11), silhouette_scores, marker='o')
plt.xlabel('Number of Clusters')
plt.ylabel('Silhouette Score')
plt.title('Silhouette Score Method')
plt.show()
```



8

9

10

Silhouette Score Method

0.42

0.40

0.36

0.34

3

4

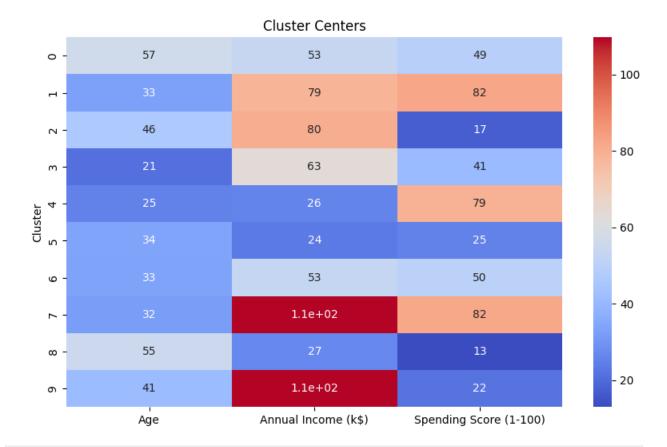
```
clusters_centers = scaler.inverse_transform(kmeans.cluster_centers_)
clusters_centers_df = pd.DataFrame(clusters_centers, columns=features)
clusters_centers_df['Cluster'] = range(kmeans.n_clusters)
data['Cluster'] = kmeans.labels_

plt.figure(figsize=(10, 6))
sns.heatmap(clusters_centers_df.set_index('Cluster'), annot=True,
cmap='coolwarm')
plt.title('Cluster Centers')
plt.show()
```

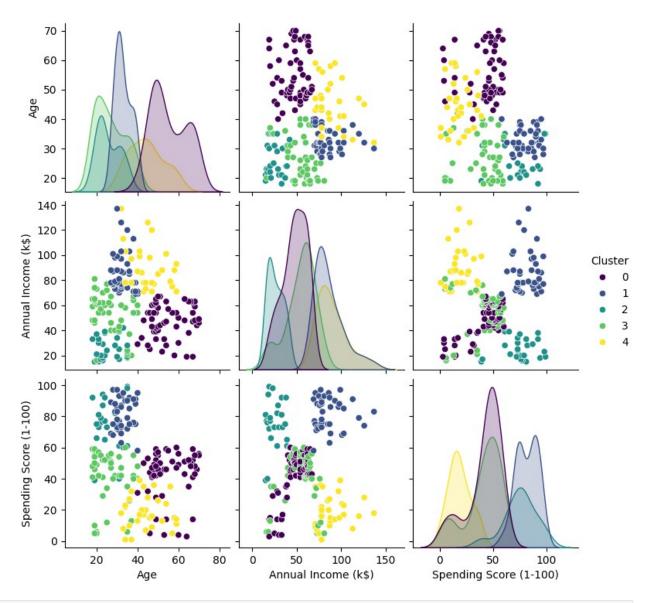
6

Number of Clusters

5



sns.pairplot(data, hue='Cluster', palette='viridis', vars=features)
plt.show()



```
plt.figure(figsize=(15, 10))
for i, feature in enumerate(features, 1):
    plt.subplot(2, 2, i)
    sns.boxplot(x='Cluster', y=feature, data=data, palette='viridis')
    plt.title(f'Distribution of {feature} in Cluster')
plt.tight_layout()
plt.show()

<ipython-input-49-3b36a88aeala>:4: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.boxplot(x='Cluster', y=feature, data=data, palette='viridis')
```

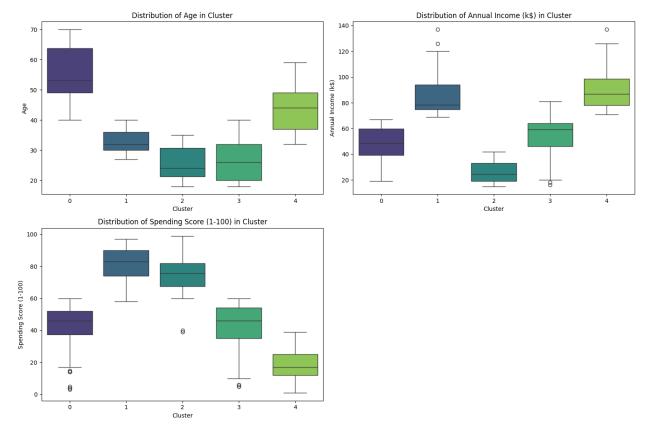
<ipython-input-49-3b36a88aea1a>:4: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.boxplot(x='Cluster', y=feature, data=data, palette='viridis')
<ipython-input-49-3b36a88aea1a>:4: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.boxplot(x='Cluster', y=feature, data=data, palette='viridis')



```
cluster_summary = data.groupby('Cluster')
[features].mean().reset index()
print(cluster summary)
data.to_csv('Clustered_customers.csv', index=False)
   Cluster
                  Age Annual Income (k$)
                                            Spending Score (1-100)
0
            55.275862
                                 47,620690
                                                         41.706897
         0
                                 86.100000
1
         1
            32.875000
                                                         81.525000
```

2	2	25.769231	26.115385	74.846154
3	3	26.733333	54.311111	40.911111
4	4	44.387097	89.774194	18.483871