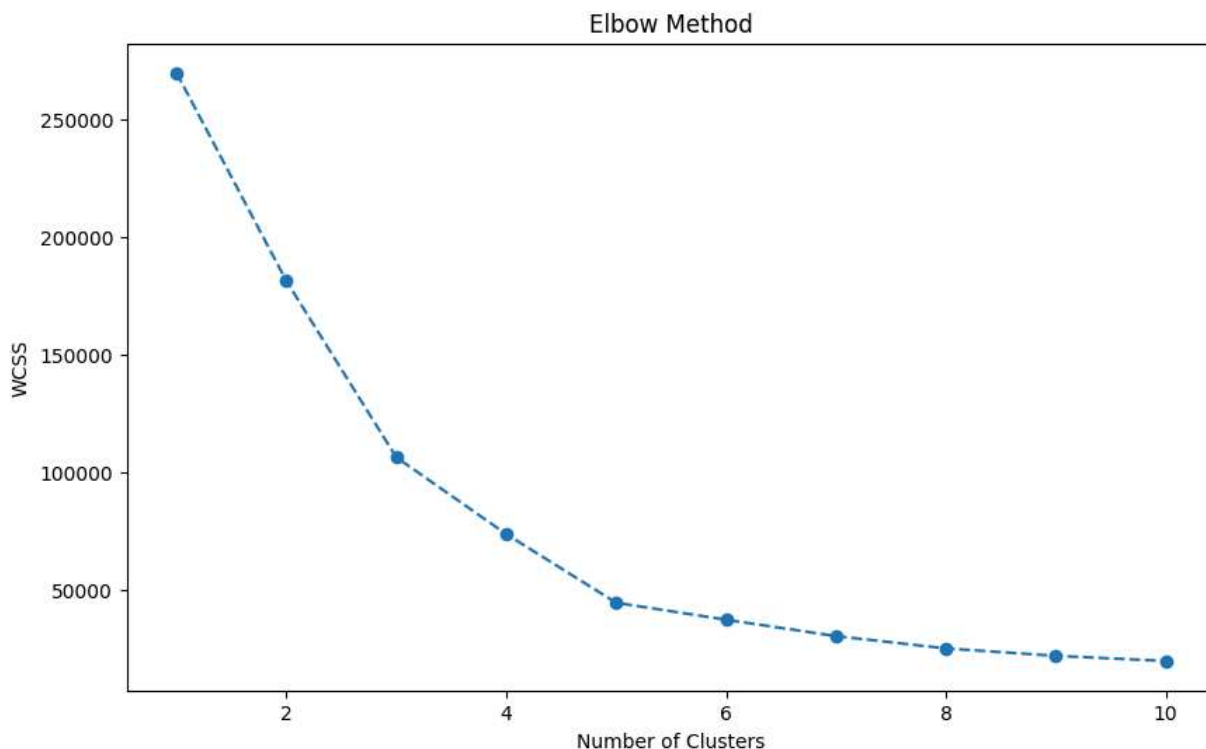


```
In [4]: # Determine the optimal number of clusters using the elbow method
wcss = []
for i in range(1, 11):
    kmeans = KMeans(n_clusters=i, init='k-means++', random_state=42)
    kmeans.fit(X)
    wcss.append(kmeans.inertia_)
```

[illegible]

```
In [5]: # Plot the elbow method graph
plt.figure(figsize=(10, 6))
plt.plot(range(1, 11), wcss, marker='o', linestyle='--')
plt.title('Elbow Method')
plt.xlabel('Number of Clusters')
```

```
plt.ylabel('WCSS')
plt.show()
```



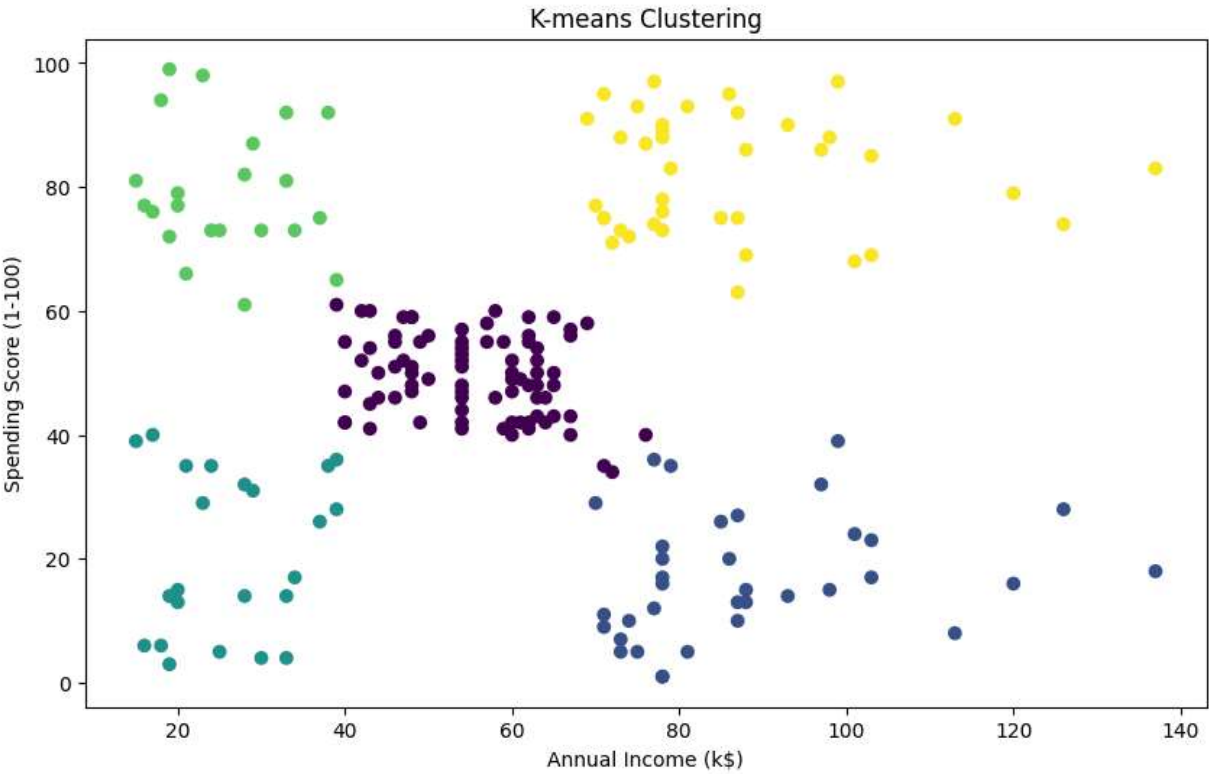
```
In [6]: # Based on the elbow method, Let's choose 5 clusters and perform K-means clustering
kmeans = KMeans(n_clusters=5, init='k-means++', random_state=42)
kmeans.fit(X)
```

/usr/local/lib/python3.10/dist-packages/sklearn/cluster/\_kmeans.py:870: FutureWarning: The default value of `n\_init` will change from 10 to 'auto' in 1.4. Set the value of `n\_init` explicitly to suppress the warning

```
Out[6]: KMeans
KMeans(n_clusters=5, random_state=42)
```

```
In [7]: # Add cluster labels to the dataframe
data['Cluster'] = kmeans.labels_
```

```
In [8]: # Plot the clusters
plt.figure(figsize=(10, 6))
plt.scatter(data['Annual Income (k$)'], data['Spending Score (1-100)'], c=data['Cluster'])
plt.title('K-means Clustering')
plt.xlabel('Annual Income (k$)')
plt.ylabel('Spending Score (1-100)')
plt.show()
```



In [ ]:

In [ ]: