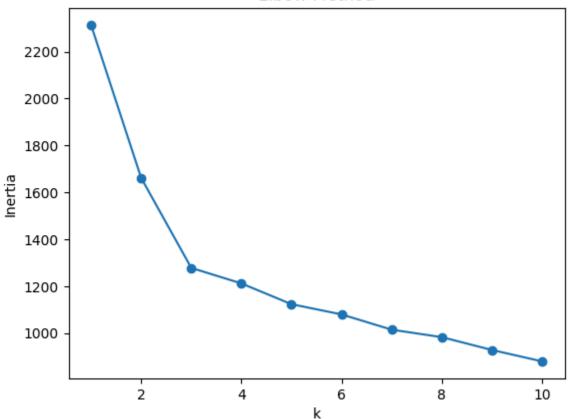


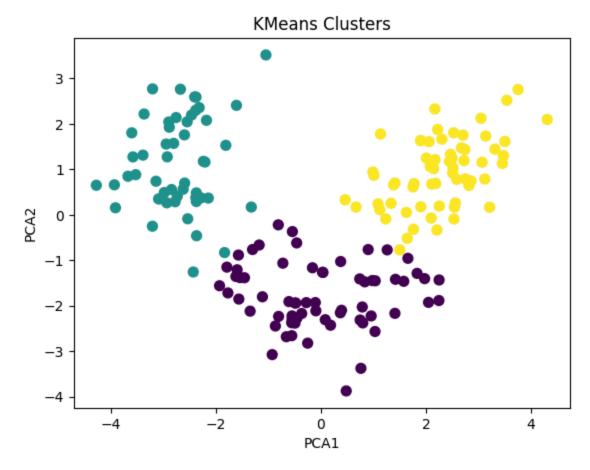
```
In [1]: import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
        from sklearn.preprocessing import StandardScaler
        from sklearn.decomposition import PCA
        from sklearn.cluster import KMeans, SpectralClustering
        from sklearn.metrics import silhouette score
        url = "https://archive.ics.uci.edu/ml/machine-learning-databases/wine/wine.dat
        col names = ["Class", "Alcohol", "Malic acid", "Ash", "Alcalinity of ash", "Magnesi
        df = pd.read csv(url, header=None, names=col names)
        X = df.drop("Class", axis=1).copy()
        scaler = StandardScaler()
        X scaled = scaler.fit transform(X)
        inertias = []
        K range = range(1, 11)
        for k in K range:
            km = KMeans(n clusters=k, random state=42)
            km.fit(X scaled)
            inertias.append(km.inertia )
        plt.plot(K range, inertias, 'o-')
        plt.xlabel("k")
        plt.ylabel("Inertia")
        plt.title("Elbow Method")
        plt.show()
        k \text{ opt} = 3
        kmeans = KMeans(n clusters=k opt, random state=42)
        labels kmeans = kmeans.fit predict(X scaled)
        sil kmeans = silhouette score(X scaled, labels kmeans)
        print("KMeans silhouette:", sil kmeans)
        pca = PCA(n components=2)
        X pca = pca.fit transform(X scaled)
        plt.scatter(X pca[:,0], X pca[:,1], c=labels kmeans, cmap='viridis', s=50)
        plt.title("KMeans Clusters")
        plt.xlabel("PCA1")
        plt.ylabel("PCA2")
        plt.show()
        base clusterings = []
        k \text{ values} = [2, 3, 4, 5]
        for k in k values:
            km2 = KMeans(n clusters=k, random state=42)
            labels2 = km2.fit predict(X scaled)
            base clusterings.append(labels2)
        base clusterings = np.array(base clusterings)
        n samples = X scaled.shape[0]
        co assoc = np.zeros((n samples, n samples))
```

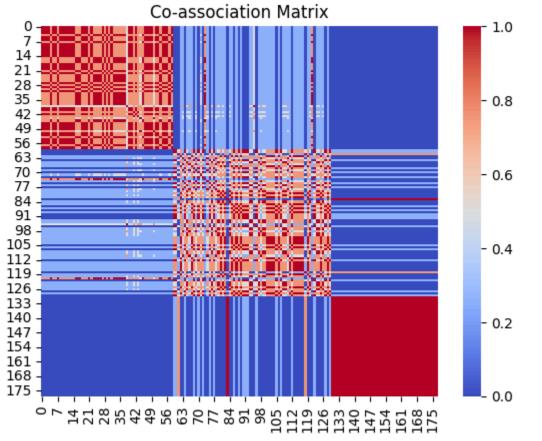
```
n runs = base clusterings.shape[0]
for run in range(n runs):
   labels = base clusterings[run]
    for i in range(n samples):
        for j in range(n samples):
            if labels[i] == labels[j]:
                co_assoc[i, j] += 1
co_assoc = co_assoc / n_runs
sns.heatmap(co assoc, cmap="coolwarm")
plt.title("Co-association Matrix")
plt.show()
spectral = SpectralClustering(n clusters=k opt, affinity='precomputed', random
labels ensemble = spectral.fit predict(co assoc)
sil ens = silhouette score(X scaled, labels ensemble)
print("Ensemble silhouette:", sil ens)
plt.scatter(X pca[:,0], X pca[:,1], c=labels ensemble, cmap='plasma', s=50)
plt.title("Ensemble Clusters")
plt.xlabel("PCA1")
plt.ylabel("PCA2")
plt.show()
print("Comparison:\nKMeans:", sil kmeans, "\nEnsemble:", sil ens)
```

Elbow Method



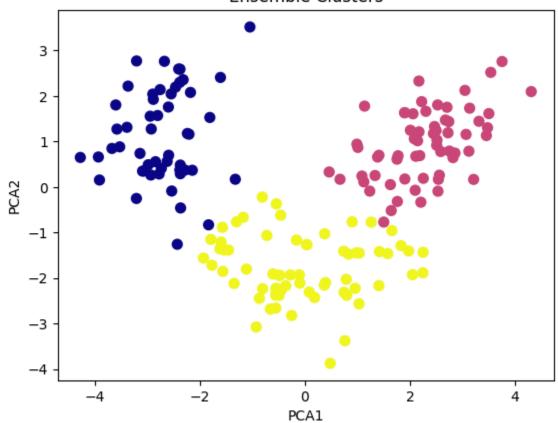
KMeans silhouette: 0.2848589191898987





Ensemble silhouette: 0.2848589191898987

Ensemble Clusters



Comparison:

KMeans: 0.2848589191898987 Ensemble: 0.2848589191898987