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import numpy as np
import pandas as pd
from sklearn.datasets import load_iris
from sklearn.preprocessing import StandardScaler
from sklearn.cluster import KMeans
from sklearn.decomposition import PCA
from sklearn.metrics import silhouette_score
import matplotlib.pyplot as plt
import seaborn as sns

iris = load_iris()
X = iris.data
y_true = iris.target
feature_names = iris.feature_names

scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)

silhouette_scores = []
K = range(2, 10)
for k in K:
    kmeans = KMeans(n_clusters=k, random_state=42, n_init=10)
    kmeans.fit(X_scaled)
    score = silhouette_score(X_scaled, kmeans.labels_)
    silhouette_scores.append(score)

plt.plot(K, silhouette_scores, 'bo-')
plt.title("Silhouette Method for Optimal k")
plt.xlabel("Number of Clusters (k)")
plt.ylabel("Silhouette Score")
plt.show()

best_k = K[np.argmax(silhouette_scores)]
print("Best number of clusters (k):", best_k)

final_kmeans = KMeans(n_clusters=best_k, random_state=42, n_init=10)
final_kmeans.fit(X_scaled)

labels = final_kmeans.labels_
centers = final_kmeans.cluster_centers_
unique, counts = np.unique(labels, return_counts=True)

print("\nCluster Labels:", np.unique(labels))
print("Cluster Sizes:", dict(zip(unique, counts)))
print("\nCluster Centers (standardized features):\n", centers)

comparison = pd.DataFrame({'True Label': y_true, 'Cluster': labels})
print("\nCluster Mapping to True Species:\n")

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print(pd.crosstab(comparison['Cluster'], comparison['True Label']))

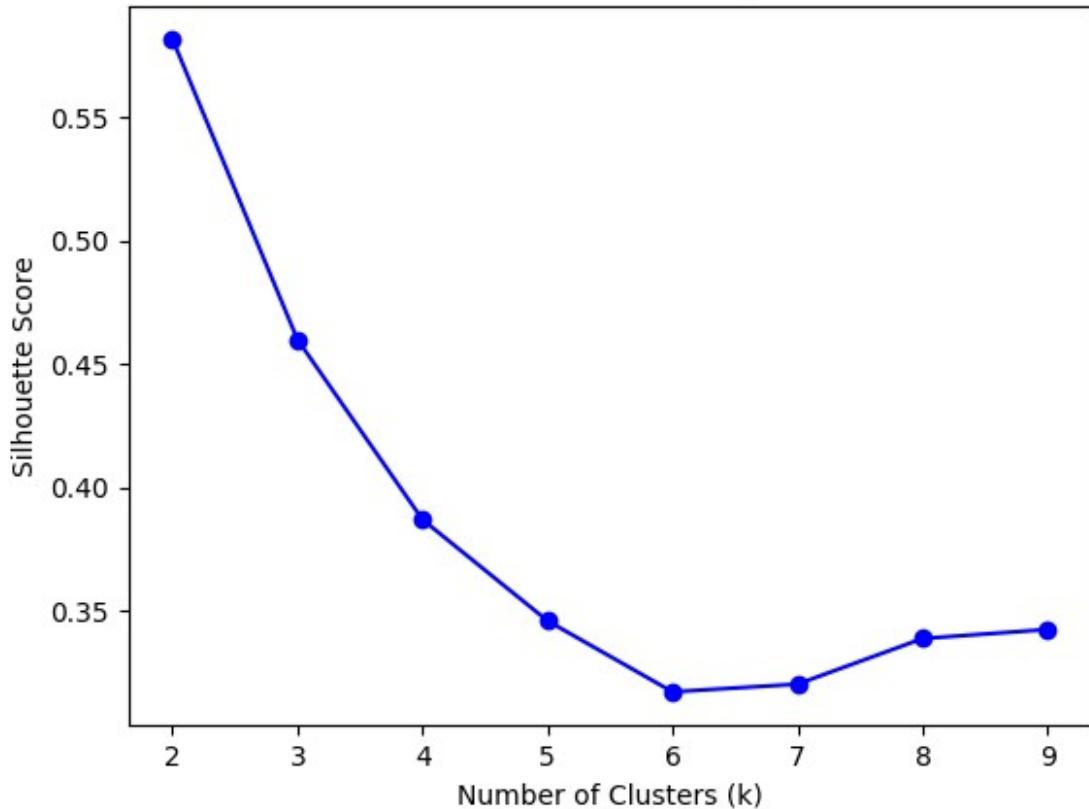
pca = PCA(n_components=2)
X_pca = pca.fit_transform(X_scaled)
centers_pca = pca.transform(centers)

plt.figure(figsize=(8, 6))
sns.scatterplot(x=X_pca[:, 0], y=X_pca[:, 1], hue=labels,
palette="Set1", s=60, alpha=0.8)
plt.scatter(centers_pca[:, 0], centers_pca[:, 1], c='black', s=200,
marker='X', label='Cluster Centers')
plt.title("K-Means Clustering with PCA (2D Projection)")
plt.xlabel("Principal Component 1")
plt.ylabel("Principal Component 2")
plt.legend()
plt.show()

print("\nCluster Size Analysis:")
for i in unique:
    print(f"Cluster {i}: {counts[i]} samples")

df_final = pd.DataFrame(X, columns=feature_names)
df_final['Cluster'] = labels
print("\nFinal Dataset with Cluster Labels:\n")
print(df_final.head())
```

Silhouette Method for Optimal k



Best number of clusters (k): 2

Cluster Labels: [0 1]

Cluster Sizes: {np.int32(0): np.int64(100), np.int32(1): np.int64(50)}

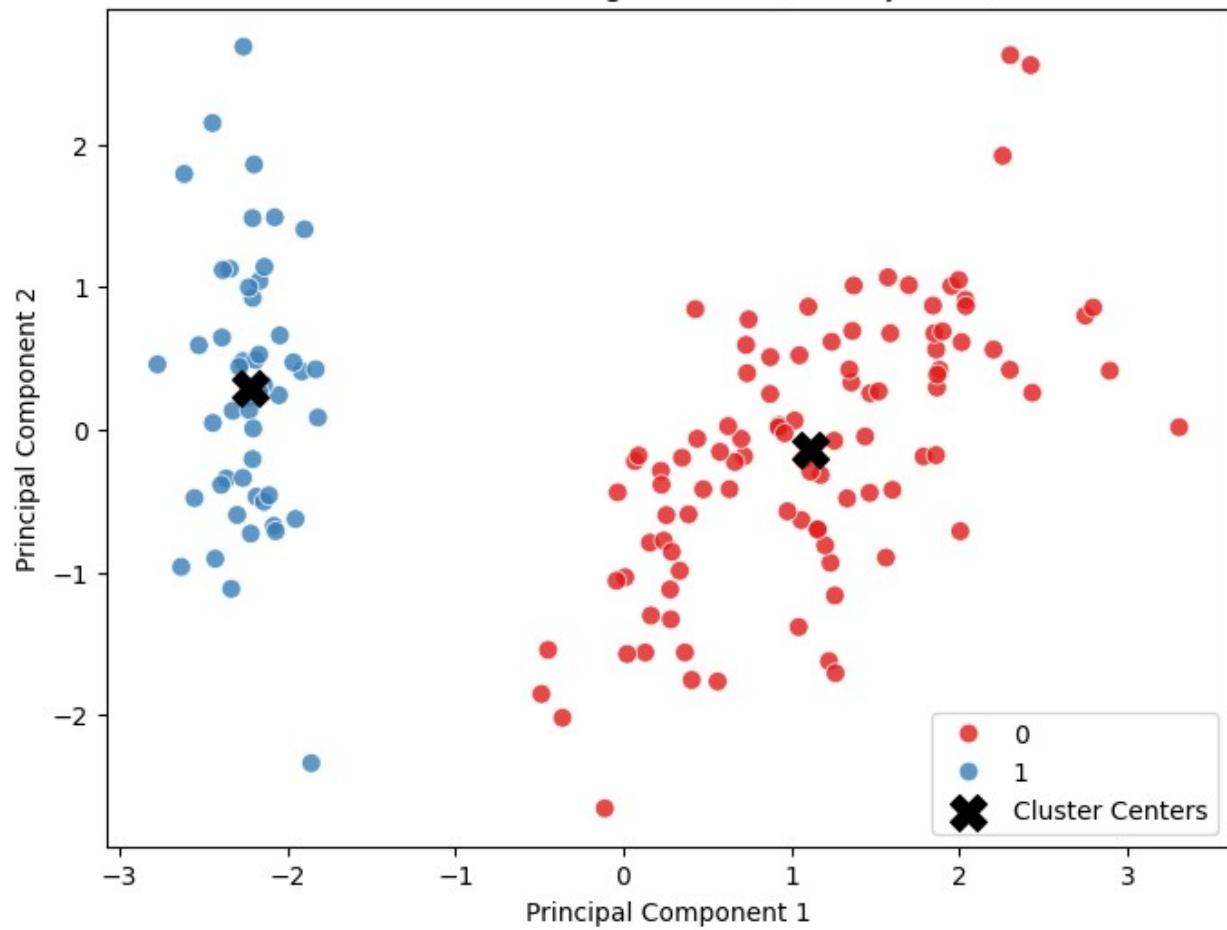
Cluster Centers (standardized features):

[[0.50728948 -0.42663134 0.65249366 0.62744675]
[-1.01457897 0.85326268 -1.30498732 -1.25489349]]

Cluster Mapping to True Species:

True Label	0	1	2
Cluster			
0	0	50	50
1	50	0	0

K-Means Clustering with PCA (2D Projection)



Cluster Size Analysis:
Cluster 0: 100 samples
Cluster 1: 50 samples

Final Dataset with Cluster Labels:

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)
(cm) \				
0	5.1		3.5	1.4
0.2				
1	4.9		3.0	1.4
0.2				
2	4.7		3.2	1.3
0.2				
3	4.6		3.1	1.5
0.2				
4	5.0		3.6	1.4
0.2				

	Cluster
0	1
1	1
2	1
3	1
4	1