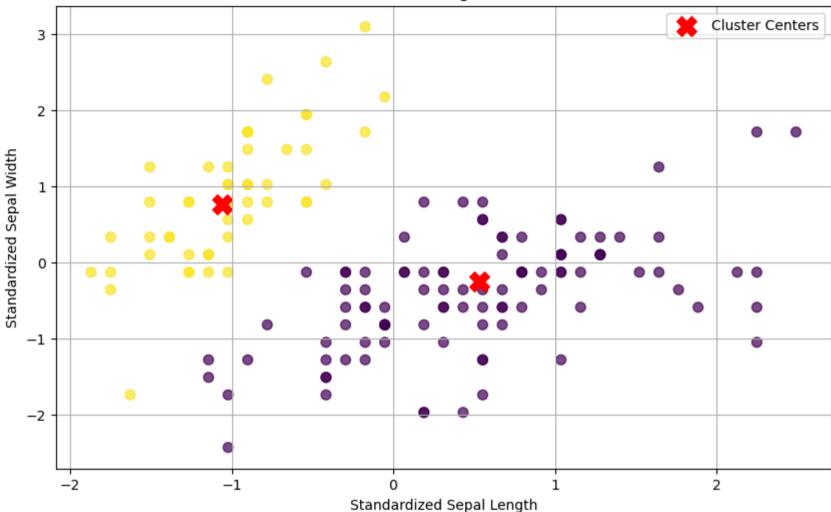
```
In [8]: import numpy as np
         import matplotlib.pyplot as plt
         from sklearn import datasets
         from sklearn.cluster import MeanShift
         from sklearn.preprocessing import StandardScaler
         from sklearn.metrics import adjusted_rand_score, normalized_mutual_info_score, silhouette_score
In [10]: | iris = datasets.load iris()
         X = iris.data # Features (sepal length, sepal width, petal length, petal width)
         v true = iris.target # Actual labels (species)
In [11]: scaler = StandardScaler()
         X_scaled = scaler.fit_transform(X)
In [12]: mean shift = MeanShift(bandwidth=1.5) # You can adjust the bandwidth parameter
         mean shift.fit(X scaled)
Out[12]: MeanShift(bandwidth=1.5)
In [13]: labels = mean_shift.labels_
         cluster centers = mean shift.cluster centers
In [14]: | n_clusters = len(np.unique(labels))
         print(f"Number of clusters found: {n clusters}")
         Number of clusters found: 2
```

```
In [7]: plt.figure(figsize=(10, 6))
# Scatter plot of the first two features (sepal length and sepal width)
plt.scatter(X_scaled[:, 0], X_scaled[:, 1], c=labels, cmap='viridis', s=50, alpha=0.7)
plt.scatter(cluster_centers[:, 0], cluster_centers[:, 1], c='red', s=200, marker='X', label='Cluster Centers'
plt.title('Mean Shift Clustering on Iris Dataset')
plt.xlabel('Standardized Sepal Length')
plt.ylabel('Standardized Sepal Width')
plt.legend()
plt.grid()
plt.show()
```

## Mean Shift Clustering on Iris Dataset



```
In [15]: ari = adjusted_rand_score(y_true, labels)
nmi = normalized_mutual_info_score(y_true, labels)
silhouette = silhouette_score(X_scaled, labels)
```

```
In [16]: # Print the evaluation metrics
    print(f"Adjusted Rand Index (ARI): {ari:.3f}")
    print(f"Normalized Mutual Information (NMI): {nmi:.3f}")
    print(f"Silhouette Score: {silhouette:.3f}")

Adjusted Rand Index (ARI): 0.568
    Normalized Mutual Information (NMI): 0.734
    Silhouette Score: 0.582
In []:
```