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In [1]: print(__doc__)
        import numpy as np
        from sklearn.cluster import DBSCAN
        from sklearn import metrics
        from sklearn.datasets import make_blobs
        from sklearn.preprocessing import StandardScaler
       Automatically created module for IPython interactive environment
In [2]: # Generate sample data
        centers = [[1, 1], [-1, -1], [1, -1]]
        X, labels_true = make_blobs(n_samples=750, centers=centers, cluster_std=0.4,
                                    random_state=0)
        X = StandardScaler().fit_transform(X)
In [6]: # Compute DBSCAN
        db = DBSCAN(eps=0.3, min_samples=10).fit(X)
        core_samples_mask = np.zeros_like(db.labels_, dtype=bool)
        core_samples_mask[db.core_sample_indices_] = True
        labels = db.labels_
        # Number of clusters in labels, ignoring noise if present.
        n_clusters_ = len(set(labels)) - (1 if -1 in labels else 0)
        n_noise_ = list(labels).count(-1)
        print('Estimated number of clusters: %d' % n_clusters_)
        print('Estimated number of noise points: %d' % n_noise_)
        print("Homogeneity: %0.3f" % metrics.homogeneity_score(labels_true, labels))
        print("Completeness: %0.3f" % metrics.completeness_score(labels_true, labels))
        print("V-measure: %0.3f" % metrics.v_measure_score(labels_true, labels))
        print("Adjusted Rand Index: %0.3f"
              % metrics.adjusted_rand_score(labels_true, labels))
        print("Adjusted Mutual Information: %0.3f"
              % metrics.adjusted_mutual_info_score(labels_true, labels))
        print("Silhouette Coefficient: %0.3f"
              % metrics.silhouette_score(X, labels))
       Estimated number of clusters: 3
       Estimated number of noise points: 18
      Homogeneity: 0.953
       Completeness: 0.883
      V-measure: 0.917
       Adjusted Rand Index: 0.952
       Adjusted Mutual Information: 0.916
       Silhouette Coefficient: 0.626
In [8]: # Plot result
        import matplotlib.pyplot as plt
        # Black removed and is used for noise instead.
        unique_labels = set(labels)
        colors = [plt.cm.Spectral(each)
                  for each in np.linspace(0, 1, len(unique_labels))]
        for k, col in zip(unique_labels, colors):
            if k == -1:
                # Black used for noise.
                col = [0, 0, 0, 1]
            class_member_mask = (labels == k)
            xy = X[class_member_mask & core_samples_mask]
            plt.plot(xy[:, 0], xy[:, 1], 'o', markerfacecolor=tuple(col),
                     markeredgecolor='k', markersize=14)
            xy = X[class_member_mask & ~core_samples_mask]
            plt.plot(xy[:, 0], xy[:, 1], 'o', markerfacecolor=tuple(col),
                     markeredgecolor='k', markersize=6)
        plt.title('Estimated number of clusters: %d' % n_clusters_)
        plt.show()
```

Estimated number of clusters: 3

