

A comparison of the clustering algorithms in scikit-learn

https://scikit-learn.org/stable/modules/clustering.html#

#### Datasets

sklearn.datasets.load\_di gits

– sklearn.datasets.fetch\_2Onewsgroups

Load and return the digits dataset (classification).

Each datapoint is a 8x8 image of a digit.

Classes	10	
Samples per class	~180	
Samples total	1797	
Dimensionality	64	
Features	integers 0-16	

Load the filenames and data from the 20 newsgroups dataset (classification).

Download it if necessary.

20
18846
1
text

• 测试sklearn中以下聚类算法在以上两个数据集上的聚类效果。

Method name	Parameters	Scalability	Usecase	Geometry (metric used)
K-Means	number of clusters	Very large n_samples, medium n_clusters with MiniBatch code	General-purpose, even cluster size, flat geometry, not too many clusters	Distances between points
Affinity propagation	damping, sample preference	Not scalable with n_samples	Many clusters, uneven cluster size, non-flat geometry	Graph distance (e.g. nearest-neighbor graph)
Mean-shift	bandwidth	Not scalable with n_samples	Many clusters, uneven cluster size, non-flat geometry	Distances between points
Spectral clustering	number of clusters	Medium n_samples, small n_clusters	Few clusters, even cluster size, non-flat geometry	Graph distance (e.g. nearest-neighbor graph)
Ward hierarchical clustering	number of clusters	Large n_samples and n_clusters	Many clusters, possibly connectivity constraints	Distances between points
Agglomerative clustering	number of clusters, linkage type, distance	Large n_samples and n_clusters	Many clusters, possibly connectivity constraints, non Euclidean distances	Any pairwise distance
DBSCAN	neighborhood size	Very large n_samples, medium n_clusters	Non-flat geometry, uneven cluster sizes	Distances between nearest points
Gaussian mixtures	many	Not scalable	Flat geometry, good for density estimation	Mahalanobis distances to centers

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#### Evaluation

- labels\_true and labels\_pred
  - >>> from sklearn import metrics
  - >>> labels\_true = [0, 0, 0, 1, 1, 1]
  - >>> labels pred = [0, 0, 1, 1, 2, 2]
- Normalized Mutual Information (NMI)
  - >>> metrics.normalized\_mutual\_info\_score(labels\_true, labels\_pred)
- Homogeneity: each cluster contains only members of a single class
  - >>> metrics.homogeneity\_score(labels\_true, labels\_pred)
- Completeness: all members of a given class are assigned to the same cluster
  - >>> metrics.completeness\_score(labels\_true, labels\_pred)

#### Examples

- A demo of K-Means clustering on the handwritten digits data
  - https://scikitlearn.org/stable/auto\_examples/cluster/plot kmeans\_digits.html#sphx-glr-autoexamples-cluster-plot-kmeans-digits-py
- Clustering text documents using kmeans
  - https://scikitlearn.org/stable/auto\_examples/text/plot\_do cument\_clustering.html#sphx-glr-autoexamples-text-plot-document-clustering-py

K-means clustering on the digits dataset (PCA-reduced data)
Centroids are marked with white cross

