## 作业六

编程如下数据聚类: 双层正方形

# 导库与全局设置

```
import matplotlib.pyplot as plt
import numpy as np
from sklearn.cluster import KMeans, DBSCAN
```

```
plt.rcParams['font.sans-serif'] = ["SimHei"]
plt.rcParams["axes.unicode_minus"] = False
```

## 生成双层正方形数据

```
a = np.arange(1, 10, 0.01)
b = np.arange(3, 8, 0.01)
w = np.zeros((5600, 3))
# 外层正方形点
w[:900, 0] = a
w[:900, 1] = 1
w[900:1800, 0] = 1
w[900:1800, 1] = a
w[1800:2700, 0] = a
w[1800:2700, 1] = 10
w[2700:3600, 0] = 10
w[2700:3600, 1] = a
# 内层正方形点
w[3600:4100, 0] = b
w[3600:4100, 1] = 3
w[4100:4600, 0] = 3
w[4100:4600, 1] = b
w[4600:5100, 0] = b
w[4600:5100, 1] = 8
w[5100:, 0] = 8
w[5100:, 1] = b
```

```
w[3600:, 2] = 1
```

## K-Means 聚类

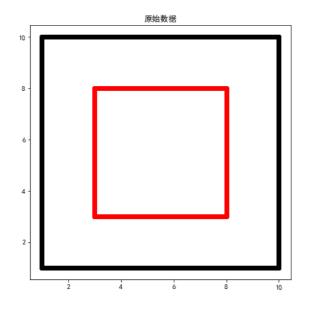
#### 参数说明

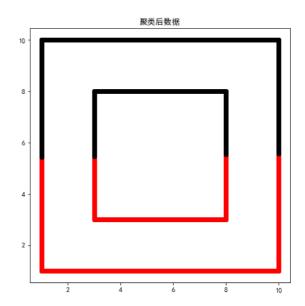
n\_clusters:聚类个数

random\_state: 控制参数随机性

```
cluster = KMeans(n_clusters=2, random_state=0)
y = cluster.fit_predict(w)
```

```
colors = ['black', 'red']
plt.figure(figsize=(15, 15))
plt.subplot(2, 2, 1)
for i in range(len(w)):
    plt.scatter(w[i][0], w[i][1],
color=colors[int(w[i][2])])
    plt.title("原始数据")
plt.subplot(2, 2, 2)
for i in range(len(y)):
    plt.scatter(w[i][0], w[i][1],
color=colors[y[i]])
    plt.title("聚类后数据")
```





### DBSCAN 聚类

#### 参数说明

eps:  $\epsilon$ -邻域的距离阈值,和样本距离超过 $\epsilon$ 的样本点不在 $\epsilon$ -邻域内,默认值是0.5。

min\_samples:形成高密度区域的最小点数。作为核心点的话邻域(即以其为圆心,eps为半径的圆,含圆上的点)中的最小样本数(包括点本身)。

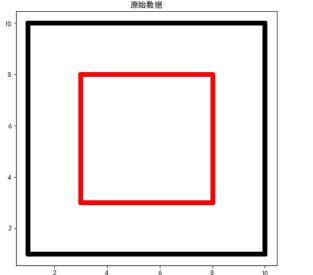
若y=-1,则为异常点。

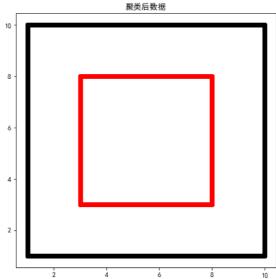
由于DBSCAN生成的类别不确定,因此定义一个函数用来筛选出符合指定类别的最合适的参数。 合适的标准是异常点个数最少。

```
# 筛选参数
def search_best_parameter(N_clusters, X):
   min_outliners = 999
   best_eps = 0
   best_min_samples = 0
   # 迭代不同的eps值
   for eps in np.arange(0.001, 1, 0.05):
       # 迭代不同的min_samples值
       for min_samples in range(2, 10):
           dbscan = DBSCAN(eps=eps,
min_samples=min_samples)
           # 模型拟合
           y = dbscan.fit_predict(X)
           # 统计各参数组合下的聚类个数(-1表示异常点)
           if len(np.argwhere(y == -1)) == 0:
               n_clusters = len(np.unique(y))
           else:
               n_clusters = len(np.unique(y)) - 1
           # 异常点的个数
           outliners = len([i for i in y if i ==
-11)
           if outliners < min outliners and
n_clusters == N_clusters:
               min_outliners = outliners
               best_eps = eps
               best_min_samples = min_samples
    return best_eps, best_min_samples
```

```
eps, min_samples = search_best_parameter(2, w)
dbscan = DBSCAN(eps=eps, min_samples=min_samples)
y = dbscan.fit_predict(w)
```

```
colors = ['black', 'red']
plt.figure(figsize=(15, 15))
plt.subplot(2, 2, 1)
for i in range(len(w)):
    plt.scatter(w[i][0], w[i][1],
color=colors[int(w[i][2])])
    plt.title("原始数据")
plt.subplot(2, 2, 2)
for i in range(len(y)):
    plt.scatter(w[i][0], w[i][1],
color=colors[y[i]])
    plt.title("聚类后数据")
```





### 总结

对于双层正方形数据来说,K-Means聚类方法不适合进行聚类,而采用DBSCAN方法可以取得较好的效果。