```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import random
import time
```

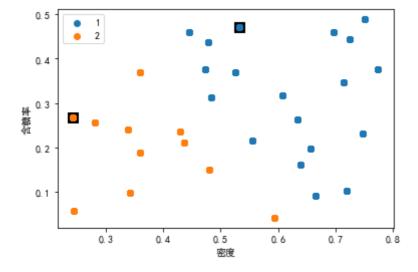
```
# 西瓜数据集4.0: 密度 含糖率 标签
data = [[0.697, 0.460, 1],
        [0.774, 0.376, 1],
        [0.634, 0.264, 1],
        [0.608, 0.318, 1],
        [0.556, 0.215, 1],
        [0.430, 0.237, 1],
        [0.481, 0.149, 1],
        [0.437, 0.211, 1],
        [0.666, 0.091, 0],
        [0.243, 0.267, 0],
        [0.245, 0.057, 0],
        [0.343, 0.099, 0],
        [0.639, 0.161, 0],
        [0.657, 0.198, 0],
        [0.360, 0.370, 0],
        [0.593, 0.042, 0],
        [0.719, 0.103, 0],
        [0.359, 0.188, 0],
        [0.339, 0.241, 0],
        [0.282, 0.257, 0],
        [0.748, 0.232, 0],
        [0.714, 0.346, 1],
        [0.483, 0.312, 1],
        [0.478, 0.437, 1],
        [0.525, 0.369, 1],
        [0.751, 0.489, 1],
        [0.532, 0.472, 1],
        [0.473, 0.376, 1],
        [0.725, 0.445, 1],
        [0.446, 0.459, 1]]
```

```
# 多维数组中创建DataFrame(二维表),需要为DataFrame赋值columns和index(默认为数字)
column = ['density', 'sugar_rate', 'label']
dataSet = pd.DataFrame(data, columns=column)
```

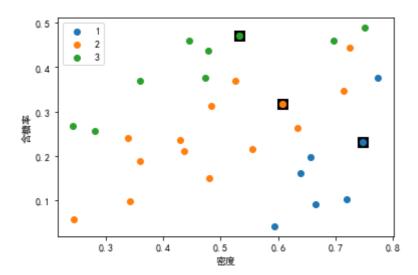
```
# 创建类K_means
class K_means(object):
   # 创建__init__方法,在面向对象编程中,给未来创建的对象所定义的进行初始化属性
   def __init__(self, k, data, loop_times, error):
       self.k = k
       self.data = data
       self.loop_times = loop_times
       self.error = error
   def distance(self, p1, p2):
       # linalg=linear (线性) +algebra (代数), norm则表示范数
       # 求p = 2 时的闵可夫斯基距离,即欧氏距离
       return np.linalg.norm(np.array(p1) - np.array(p2))
   def fitting(self):
       time1 = time.perf_counter() # 返回性能计数器的值(以分秒为单位),表示程序开始运
行到调用这个语句所经历的时间
       mean_vectors = random.sample(self.data, self.k) # 随机选取k个初始样本
       initial_main_vectors = mean_vectors
       for vec in mean_vectors :
          plt.scatter(vec[0], vec[1], s=100, color = 'black', marker='s') # I
出初始聚类中心,以黑色正方形(square)表示
       times = 0
       # map(),高阶函数,它接收一个函数 f 和一个 list,并通过把函数 f 依次作用在 list 的
每个元素上,得到一个新的 list 并返回
       # lambda:返回可调用的函数对象,通常是在需要一个函数,但又不想命名一个函数时使用,
lambda x : [x] 表示输入x, 输出为[x]
       clusters = list(map((lambda x:[x]), mean_vectors))
       while times < self.loop_times:</pre>
          change_flag = 1 # 标记簇均值向量是否改变
          for sample in self.data:
              dist = []
              for vec in mean_vectors:
                 dist.append(self.distance(vec, sample)) # 计算样本到每个聚类中
心的距离
              clusters[dist.index(min(dist))].append(sample) # 找到离该样本最近的
聚类中心,并将它放入该簇
          new_mean_vectors = []
          for c,v in zip(clusters, mean_vectors): # zip()将两个对象中对应的元素打
包成一个个元组, 然后返回由这些元组组成的列表
              cluster_num = len(c)
              cluster_array = np.array(c)
              new_mean_vector = sum(cluster_array) / cluster_num # 计算出新的聚
类簇均值向量
              mean\_vector = np.array(v)
              # np.divide和np.true_divide结果一样 (python3.7.2), np.floor_divide
只保留整数结果
              # all(iterable):如果iterable(元组或者列表)的所有元素不为0、False或者
iterable为空, all(iterable)返回True, 否则返回False
```

```
if all(np.true_divide((new_mean_vector - mean_vector),
mean_vector) < np.array([self.error, self.error])):</pre>
                   new_mean_vectors.append(mean_vector) # 均值向量未改变
                   change_flag = 0
               else:
                   # dataFrame转List(), 括号不能忘
                   new_mean_vectors.append(new_mean_vector.tolist()) # 均值向量
发生改变
           if change_flag == 1:
               mean_vectors = new_mean_vectors
           else:
               break
           times += 1
       time2 = time.perf_counter()
       # str.format(), 基本语法是通过 {} 和 : 来代替以前的 %
       print ('本次选取的{}个初始向量为{}'.format(self.k, initial_main_vectors))
       print ('共进行{}轮'.format(times))
       print ('共耗时{:.2f}s'.format(time2 - time1)) # 取2位小数
       for cluster in clusters:
           x = list(map(lambda arr: arr[0], cluster))
           y = list(map(lambda arr: arr[1], cluster))
           plt.scatter(x, y, marker = 'o', label = clusters.index(cluster)+1)
       plt.xlabel('密度')
       plt.ylabel('含糖率')
       plt.rcParams['font.sans-serif'] = ['SimHei'] # 用来正常显示中文标签
       plt.legend(loc='upper left')
       plt.show()
```

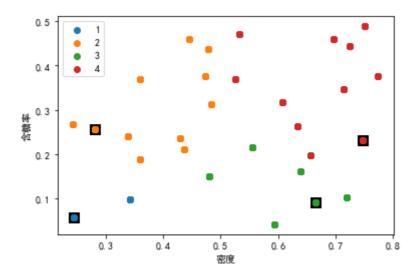
```
本次选取的2个初始向量为[[0.532, 0.472], [0.243, 0.267]]
共进行276轮
共耗时0.90s
```



本次选取的3个初始向量为[[0.748, 0.232], [0.608, 0.318], [0.532, 0.472]] 共进行0轮 共耗时0.02s

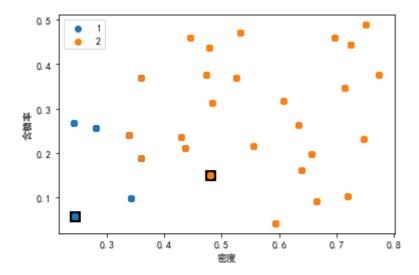


本次选取的4个初始向量为[[0.245, 0.057], [0.282, 0.257], [0.666, 0.091], [0.748, 0.232]] 共进行508轮 共耗时3.02s

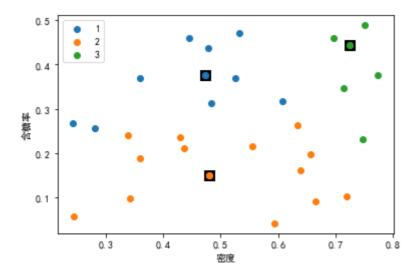


```
for i in [2, 3, 4]:
    # 调用K_means, 执行方法fitting()
    k_means = K_means(i, dataSet[['density', 'sugar_rate']].values.tolist(),
1000, 0.0000001)
    k_means.fitting()
```

本次选取的2个初始向量为[[0.245, 0.057], [0.481, 0.149]] 共进行1000轮 共耗时11.07s



本次选取的3个初始向量为[[0.473, 0.376], [0.481, 0.149], [0.725, 0.445]] 共进行**0**轮 共耗时**0**.09s



本次选取的4个初始向量为[[0.657, 0.198], [0.339, 0.241], [0.478, 0.437], [0.473, 0.376]] 共进行0轮 共耗时0.01s

