LR test

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[11]: import numpy as np
     import matplotlib.pyplot as plt
     import threading
     import random
     from time import time
[12]: # sigmoid 函数
     def sigmoid(x):
         return 1 / (1 + np.exp(-x))
[13]: # 从文件中读取训练数据
     def parse_train_data(filename):
         data = np.loadtxt(fname=filename, delimiter=',')
         dataMat = data[:, 0:-1]
         classLabels = data[:, -1]
         dataMat = np.insert(dataMat, 0, 1, axis=1) # 按列在每行位置 O 插入 1, 即偏置
     项,与 weights 对齐
         return dataMat, classLabels
[14]: # 从文件中读取测试数据
     def parse_test_data(filename):
         data = np.loadtxt(fname=filename, delimiter=',')
         dataMat = data[:, :]
         dataMat = np.insert(dataMat, 0, 1, axis=1) # 按列在每行位置 O 插入 1, 即偏置
     项,与 weights 对齐
         return dataMat
[15]: # 损失函数
     def loss_funtion(dataMat, classLabels, weights):
         m, n = np.shape(dataMat)
         loss = 0.0
         for i in range(m):
             logit = 0.0
             for j in range(n):
```

```
[16]: # 梯度上升
     def grad_Ascent(dataMatIn, classLabels):
         dataMatrix = np.mat(dataMatIn) # (m,n)
         labelMat = np.mat(classLabels).T
         m, n = np.shape(dataMatrix)
         weights = np.ones((n, 1)) # 列向量
         alpha = 0.01
         maxstep = 10 # 迭代次数
         eps = 0.0001 # 损失小于一个阈值返回
         count = 0
         loss_array = []
         for i in range(maxstep):
             loss = loss_funtion(dataMatrix, labelMat, weights)
             h_{t} = sigmoid(dataMatrix * weights) # g(h(x))
             e = labelMat - h_theta_x # y-q(h(x))
             new_weights = weights + alpha * dataMatrix.T * e # 迭代
             new_loss = loss_funtion(dataMatrix, labelMat, new_weights)
             loss_array.append(new_loss)
             if abs(new_loss - loss) < eps:</pre>
                 break
             else:
                 weights = new_weights
                 count += 1
         print("count is: ", count)
         print("loss is: ", loss)
         print("weights is: ", weights)
         return weights, loss_array
```

```
[17]: # 随机梯度上升
def stocGradAscent(dataMatrix, classLabels, numIter=100):
    m, n = np.shape(dataMatrix) # 返回 dataMatrix 的大小。m 为行数,n 为列数。
    weights = np.ones(n) # 参数初始化
    for j in range(numIter):
        dataIndex = list(range(m)) # 记录样本点索引
    for i in range(m):
        alpha = 4/(1.0+j+i)+0.01 # 降低 alpha 的大小,每次减小 1/(j+i)。
```

```
[18]: # 分类函数 特征向量,回归系数
def classifyVector(inX, weights):
    prob = sigmoid(sum(inX*weights))
    if prob > 0.5:
        return 1.0
    else:
        return 0.0
```

```
[20]: if __name__ == '__main__':
    start=time()
    lr()
    end=time()
    print('耗时: %d 秒' %(end-start))
```

测试集准确率为: 83.80%

耗时: 217 秒