LR test

April 27, 2021

方理楠-201270001

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[]: import numpy as np
    import matplotlib.pyplot as plt
    import threading
    import random
    from time import time
[]: # sigmoid 函数
    def sigmoid(x):
        return 1 / (1 + np.exp(-x))
[]: # 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
[]: # 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
[]: # 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):
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logit += dataMat[i, j] * weights.T[0, j] # 计算一个样本的
log-odds(logit)

propability = sigmoid(logit)

loss += classLabels[i, 0] * np.log(propability) + (

1 - classLabels[i, 0]) * np.log(1 - propability) # 损失函数
return loss
```

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[]: # 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
```

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[]: # 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)。

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randIndex = int(random.uniform(0, len(dataIndex))) # 随机选取样本
                # 选择随机选取的一个样本, 计算 h
                h = sigmoid(sum(dataMatrix[dataIndex[randIndex]]*weights))
                error = classLabels[dataIndex[randIndex]] - h # 计算误差
                weights = weights + alpha * error * \
                    dataMatrix[dataIndex[randIndex]] # 更新回归系数
                del(dataIndex[randIndex]) # 删除已经使用的样本
        return weights #返回
[]: # 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
[]: def lr():
        data_train, labels_train = parse_train_data('train_data.txt')
        train_weights = stocGradAscent(data_train, labels_train, 100)
        data test = parse test data('test data.txt')
        labels_test = np.loadtxt('answer.txt')
        correct_count = 0
        m, n = np.shape(data_test)
        for i in range(m):
            if classifyVector(data_test[i], train_weights) == labels_test[i]:
                correct_count += 1
        correct_rate = (correct_count/m)*100
        print("测试集准确率为: %.2f%%" % correct_rate)
[]: if __name__ == '__main__':
        start=time()
        lr()
        end=time()
        print('耗时: %d 秒' %(end-start))
[]:
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