

ml-01

October 30, 2018

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In [1]: # -*- coding: UTF-8 -*-
#numpyhttps://www.jianshu.com/p/a260a8c43e44
#matplotlibhttps://matplotlib.org/api/index.html

import xlrd      # xlrd xlrd.open_workbook excel
import matplotlib.pyplot as plt
# matplotlib matplotlib.pyplot plt, plt
import numpy as np
# numpy np, np

# loadData filename data.xlsnumpyarray
def loadData(filename):
    workbook = xlrd.open_workbook(filename)
    boyinfo = workbook.sheet_by_index(0)
    col_num = boyinfo.ncols
    row_num = boyinfo.nrows
    col0 = boyinfo.col_values(0)[1:]
    data = np.array(col0)
    if col_num == 1:
        return data
    else:
        for i in range(col_num-1):
            coltemp = boyinfo.col_values(i+1)[1:]
            data = np.c_[data, coltemp]
        return data

# plotData X flag: y plt, p1, p2
def plotData(X, y):
    pos = np.where(y==1)
    # np.where y == 1 pos
    neg = np.where(y==0)
    # np.where y == 0 neg
    # plt.plot y == 1 s () square, 7 red
    p1 = plt.plot(X[pos, 0], X[pos, 1], marker='s', markersize=7, color='red')[0]
    # plt.plot y == 0 o () circle, 7 green
    p2 = plt.plot(X[neg, 0], X[neg, 1], marker='o', markersize=7, color='green')[0] #
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    return p1, p2                                     # plt, p1, p2

# normalization normalization X X_norm X_norm
def normalization(X):
    Xmin = np.min(X,axis=0)#
    # np.min axis=0 Xmin
    Xmax =np.max(X,axis=0) #
    # np.max axis=0 Xmax
    Xmu =np.mean(X,axis=0) #
    # np.mean Xmu
    X_norm = (X-Xmu)/(Xmax-Xmin) #
    # (X-Xmu)/(Xmax-Xmin) [-1,1]
    return X_norm # X_norm

# plot decision boundary plotDecisionBoundaryn trainX, trainY, w, iter_num
def plotDecisionBoundary(trainX, trainY, w, iter_num = 0):
    # prepare data
    xcord1 = [];
    ycord1 = [];
    xcord2 = [];
    ycord2 = []
    # xcord1ycord1xcord2ycord2
    m, n = np.shape(trainX)
    # np.shape trainX m trainX n trainX
    for i in range(m):
        # for trainX i 012...m-1 m
        if trainY[i] == 1:
# if trainY 1 trainX trainX[i,1] trainX[i,2] xcord1 ycord1
            xcord1.append(trainX[i,1])
# append trainX trainX[i,1] xcord1 pos , positive
            ycord1.append(trainX[i,2])
# append trainX trainX[i,2] ycord1 pos , positive
        else:
            # trainY 1 trainX trainX[i,1] trainX[i,2] xcord2 ycord2
            xcord2.append(trainX[i,1])
            # append trainX trainX[i,1] xcord2 neg , negative
            ycord2.append(trainX[i,2])
            # append trainX trainX[i,2] ycord2 neg , negative
    x_min = min(trainX[:,1]) # min trainX[:,1] trainX 2 x_min
    y_min = min(trainX[:,2]) # min trainX[:,2] trainX 3 y_min
    x_max = max(trainX[:,1]) # max trainX[:,1] trainX 2 x_max
    y_max = max(trainX[:,2]) # max trainX[:,2] trainX 3 y_max

    # plot scatter & legend
    fig = plt.figure(1) # plt.figure fig
    # plt.scatter xcord1, ycord130 s () square, 'I like you'
    plt.scatter(xcord1, ycord1, s=30, c='red', marker='s', label='I like you')
    plt.scatter(xcord2,ycord2,s=30,c='green',marker='o',label="I don't like you")

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# plt.scatter xcord2, ycord230 o () circle, 'I don't like you'
#
plt.legend(loc='upper right') #
# set axis and ticks
delta_x = x_max-x_min # delta_x
delta_y = y_max-y_min # delta_y
# x_min - delta_x / 10 x_max + delta_x / 10 np.arange 1 my_x_ticks
my_x_ticks = np.arange(x_min - delta_x / 10, x_max + delta_x / 10, 1)
# y_min - delta_y / 10 y_max + delta_y / 10 np.arange 1 my_y_ticks
my_y_ticks = np.arange(y_min - delta_y / 10, y_max + delta_y / 10, 1)

plt.xticks(my_x_ticks) # plt.xticks my_x_ticks
plt.yticks(my_y_ticks) # plt.yticks my_y_ticks
# plt.axis [x_min-delta_x/10, x_max+delta_x/10] [y_min-delta_y/10, y_max+delta_y/10]
plt.axis([x_min-delta_x/10, x_max+delta_x/10, y_min-delta_y/10, y_max+delta_y/10])

# draw a line
x = np.arange(x_min-delta_x/10, x_max+delta_x/10, 0.01) # np.arange x_min - delta
y = (-w[0]-w[1]*x)/w[2] # y = (-w[0]-w[1]*x)/w[2]
plt.plot(x, y.T) # plt.plot x , y.T.T

# figure name
# 'Training ' + str(iter_num) + ' times.png' str(iter_num) iter_num png
fig_name = 'Training ' + str(iter_num) + ' times.png'
# 'Training ' + str(iter_num) + ' times.png' str(iter_num) iter_num png
plt.title(fig_name)
fig.savefig(fig_name) # fig.savefig
plt.show(fig) # plt.show
plt.close()

# sigmoid: sigmoid activation function wx sigmoid
def sigmoid(wx):
    sigmoidV = 1.0/(1.0+np.exp(-wx)) # sigmoid 1.0/(1.0+np.exp(-wx))
    return sigmoidV

# loss fuc Y_ Y
def loss(X, Y, w):
    # loss loss function X, Y, w
    m, n = np.shape(X)
    # np.shape X m X n X
    trainMat = np.mat(X)
    # np.mat X trainMat
    Y_ = [] # Y_, append
    for i in np.arange(m): # for X i 0 1 2....m-1 X m
        # append Y_ trainMat[i] w sigmoid
        Y_.append(sigmoid(trainMat[i]*w))
    m = np.shape(Y_)[0] # np.shape X np.shape(Y_)[0] X m
    sum_err = 0.0 # 0.0, sum_err sum_err
    for i in range(m): # for Y_ i 0 1 2....m-1 Y_ m

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        # sum_err sum_err Y[i]*np.log(Y_[i])+(1-Y[i])*np.log(1-Y_[i]) Cross Entro
        sum_err -= Y[i]*np.log(Y_[i])+(1-Y[i])*np.log(1-Y_[i]) #
    return sum_err/m # sum_err

# BGD
# BGD Batch Gradient DescentBGD X y,
# iter_num, alpha lr (learning rate), J w
# Batch Gradient DescentBGD W
def BGD(X, y, iter_num, alpha):
    trainMat = np.mat(X) # np.mat X trainMat
    trainY = np.mat(y).T # np.mat y trainY
    m, n = np.shape(X) # np.shape X m X n X
    w = np.ones((n,1)) # np.ones 1 n 1 w, w 1
    for i in range(iter_num): # for i 01 2....iter_num-1 iter_num
        error = sigmoid(trainMat*w)-trainY # error sigmoid
        w =w - (1.0/m)*alpha*trainMat.T*error # w , B
    return w # w

# classify classify wx 1 0
def classify(wx):
    prob = sigmoid(wx) # sigmoid(wx) prob
    if prob > 0.5: # prob 0.5 1
        return 1
    else: # prob 0.5 0
        return 0

# predict predict testX w result
def predict(testX, w):
    m, n = np.shape(testX) # np.shape testX m testX n testX
    testMat = np.mat(testX) # np.mat testX testMat
    result = [] # result, append
    for i in np.arange(m): # for testX i 01 2....m-1 testX
        # append result classify 1 0 result
        result.append(classify(float(testMat[i]*w)))
    return result # result

# Precision Precision X, Y w
def Precision(X, Y, w):
    result = predict(X, w) # predict X w result
    right_sum = 0 # 0 right_sum 1
    # for i 01 2....len(result)-1 result len(result)
    for i in range(len(result)):
        if result[i]-int(Y[i]) == 0: # if, result int(Y[i]) right_sum
            right_sum += 1 # right_sum 1
    # 1.0*right_sum/len(Y) 1.0 float
    return 1.0*right_sum/len(Y)

# python

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if __name__ == "__main__":
    # load data and visualization
    data = loadData('data.xls')
    X = data[:,2]
    y = data[:,2]

    # plot data
    plt_data = plt.figure(1)
    p1, p2 = plotData(X, y)

    #Labels and Legend
    plt.xlabel('tall')
    plt.ylabel('salary')
    # plt.legend 'I like you' "I don't like you"
    # numpoints 1 handlelength 0
    plt.legend((p1, p2), ('I like you', "I don't like you"), numpoints=1, handlelength=0)

    # show and save visualized image
    plt_data.savefig('visualization_org.jpg')
    plt.show(plt_data)
    plt.close(plt_data)

    # normalization and visualization normalization X
    X_norm = normalization(X)
    # plot data
    plt_norm = plt.figure(1)
    # plotData X_norm y plt_norm, p1_norm p2_norm
    p1_norm, p2_norm = plotData(X_norm, y)

    # Labels and Legend
    plt.xlabel('tall')
    plt.ylabel('salary')
    # plt.legend 'I like you' "I don't like you"
    # numpoints 1 handlelength 0
    plt.legend((p1_norm, p2_norm), ('I like you', "I don't like you"), numpoints=1, handlelength=0)

    # show and save visualized image
    # plt.show
    #
    plt.show(plt_norm)
    # plt.savefig 'visualization_norm.jpg' 'jpg'
    #
    figname='visualization_norm.jpg'
    plt.savefig(figname)
    # plt.close
    #
    plt.close()

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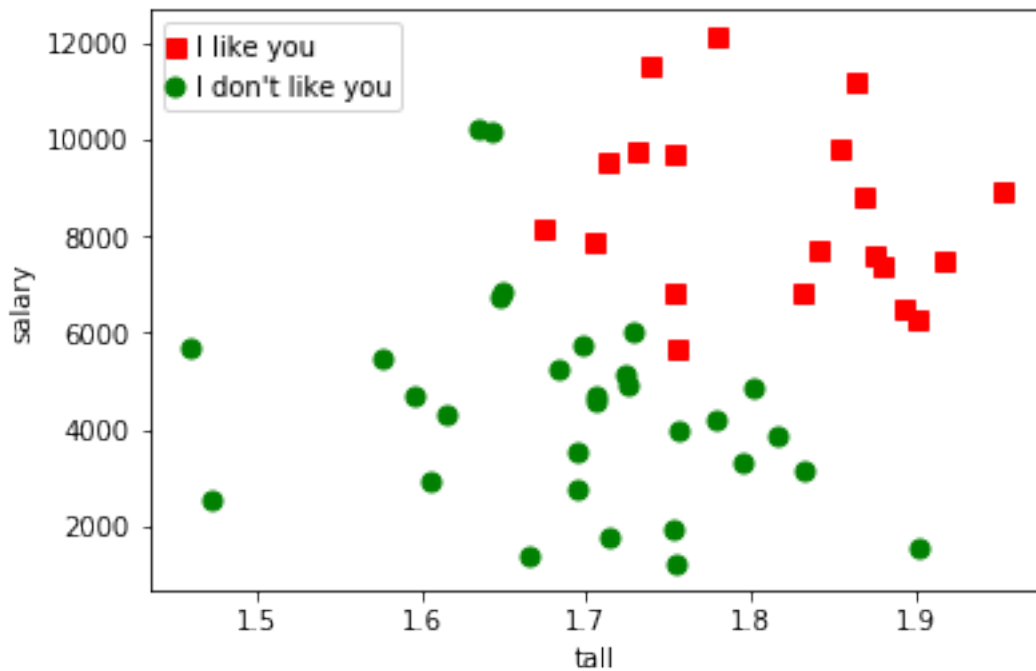
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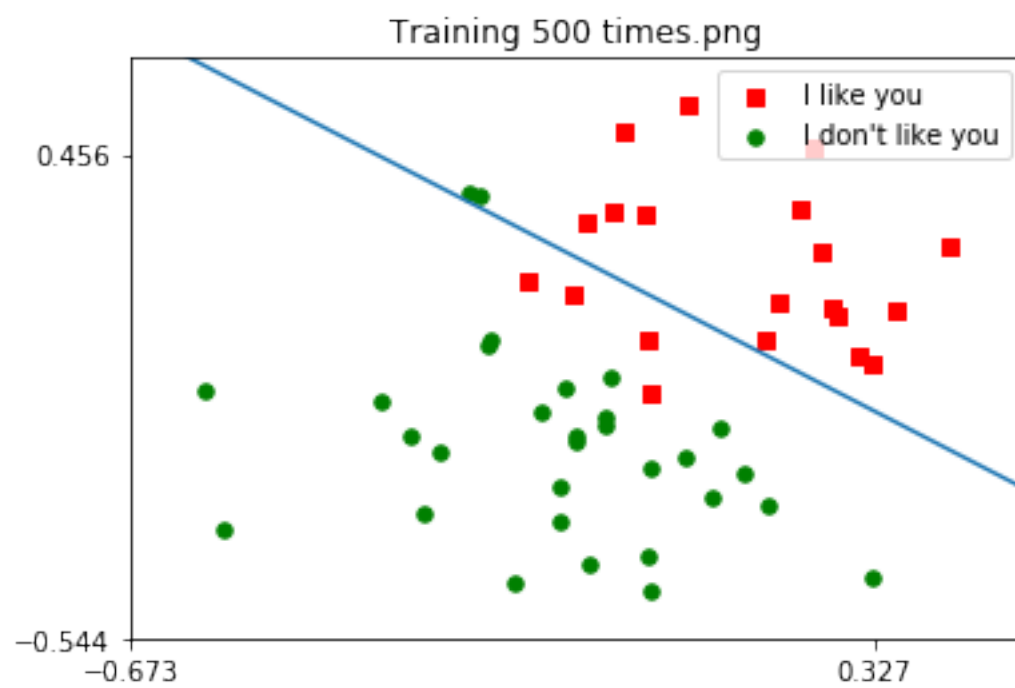
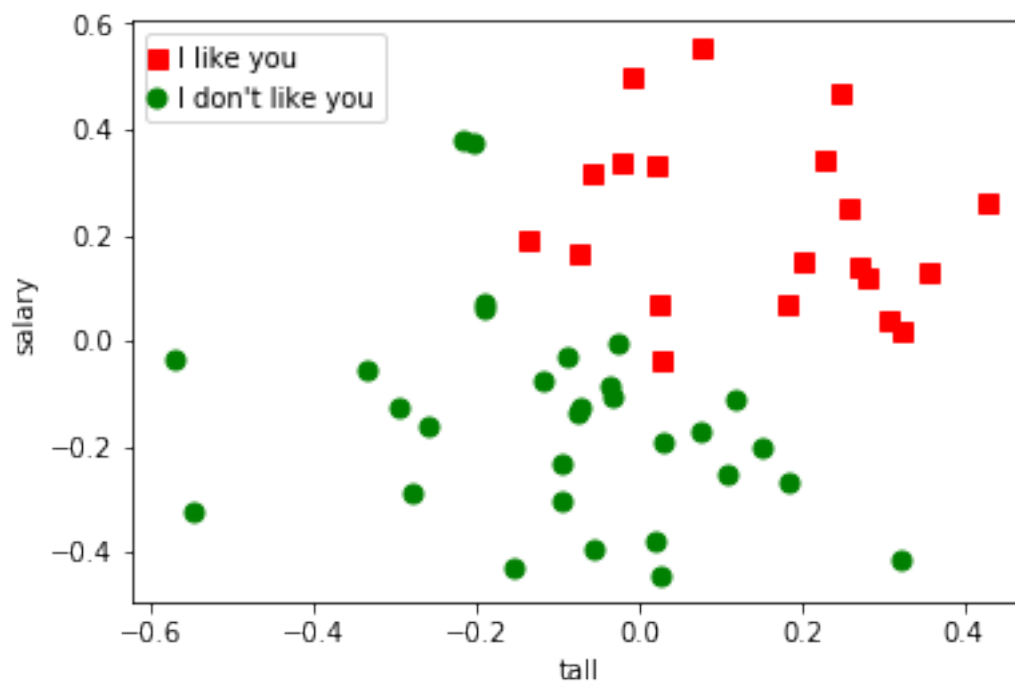
# optimizing by BSD
iter_num=500
lr=0.05
m,n = np.shape(data)
offset = np.ones((m, 1))
trainMat = np.c_[offset, X_norm]
theta=BGD(trainMat,y,iter_num,lr)

## Plot Boundary
# plotDecisionBoundary trainMat, y, theta iter_num
plotDecisionBoundary(trainMat, y, theta, iter_num)
cost = loss(trainMat, y, theta)
print('Cost theta: {0}'.format(cost))

# Compute accuracy on our training set
p = Precision(trainMat, y, theta)
print('Train Accuracy: {0}'.format(p))
print('finished!')

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Cost theta: $[[0.42092422]]$
 Train Accuracy: 0.88

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finished!
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In [ ]:
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