

# CSE 250A HW5.5

November 6, 2017

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In [1]: '''
        Performing logistic regression via GRADIENT ASCENT to classify handwritten digits
        '''

Out[1]: '\nPerforming logistic regression via GRADIENT ASCENT to classify handwritten digits\n'

In [75]: import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline

In [76]: # load files
train3_fh = 'hw5_train3.txt'
test3_fh = 'hw5_test3.txt'
train5_fh = 'hw5_train5.txt'
test5_fh = 'hw5_test5.txt'

train3 = np.loadtxt(train3_fh, dtype=int)
test3 = np.loadtxt(test3_fh, dtype=int)
train5 = np.loadtxt(train5_fh, dtype=int)
test5 = np.loadtxt(test5_fh, dtype=int)

# collect together
train = np.append(train3, train5, axis=0)
test = np.append(test3, test5, axis=0)
train_labs = [0] * train3.shape[0] + [1] * train5.shape[0]
test_labs = [0] * test3.shape[0] + [1] * test5.shape[0]

# CONSTANTS
STEPS = 5000

In [83]: # functions
def sigmoid(w, xt):
    z = np.dot(w, xt)
    return(1/(1+np.exp(-z)))

def log_likelihood(yt, w, xt):
    L = yt*np.log(sigmoid(w, xt)) + (1-yt)*np.log((1-sigmoid(w, xt)))
    return(L)
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def gradient(yt,w,xt):
    dL = np.multiply(yt-sigmoid(w, xt), xt)
    return(dL)

def learn(x_data, y_data):
    T = x_data.shape[0]
    eta = 0.02/T
    w = np.random.randint(2, size=x_data.shape[1])
    Lw_list = []
    percent_err = []
    for i in range(STEPS):
        Lw = 0
        dL_sum = [0] * x_data.shape[1]
        correct = 0
        for j in range(T):
            Lw += log_likelihood(y_data[j], w, x_data[j])
            dL_sum += gradient(y_data[j], w, x_data[j])
            if (y_data[j]==1 and sigmoid(w, x_data[j]) > 0.5) or (y_data[j]==0 and sigmoid(w, x_data[j]) < 0.5):
                correct += 1
        Lw_list.append(Lw)
        w = w + eta*dL_sum
        err = (T-correct)/T*1.0
        percent_err.append(err)
        if i%100 == 0 :
            print('iteration %d' % i)
            print('percent error=%f' % err)
    return Lw_list, w, percent_err

def predict(x_data, y_data, w):
    correct = 0
    for i in range(x_data.shape[0]):
        s = sigmoid(w,x_data[i])
        if (y_data[i]==1 and s>0.5) or (y_data[i]==0 and s<0.5):
            correct += 1
    err = (x_data.shape[0]-correct)/x_data.shape[0]*1.0
    return err

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In [42]: Lw\_train, w\_train, err\_train = learn(train, train\_labs)

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iteration 0
err=0.500000
iteration 100
err=0.246429
iteration 200
err=0.177143
iteration 300
err=0.140000

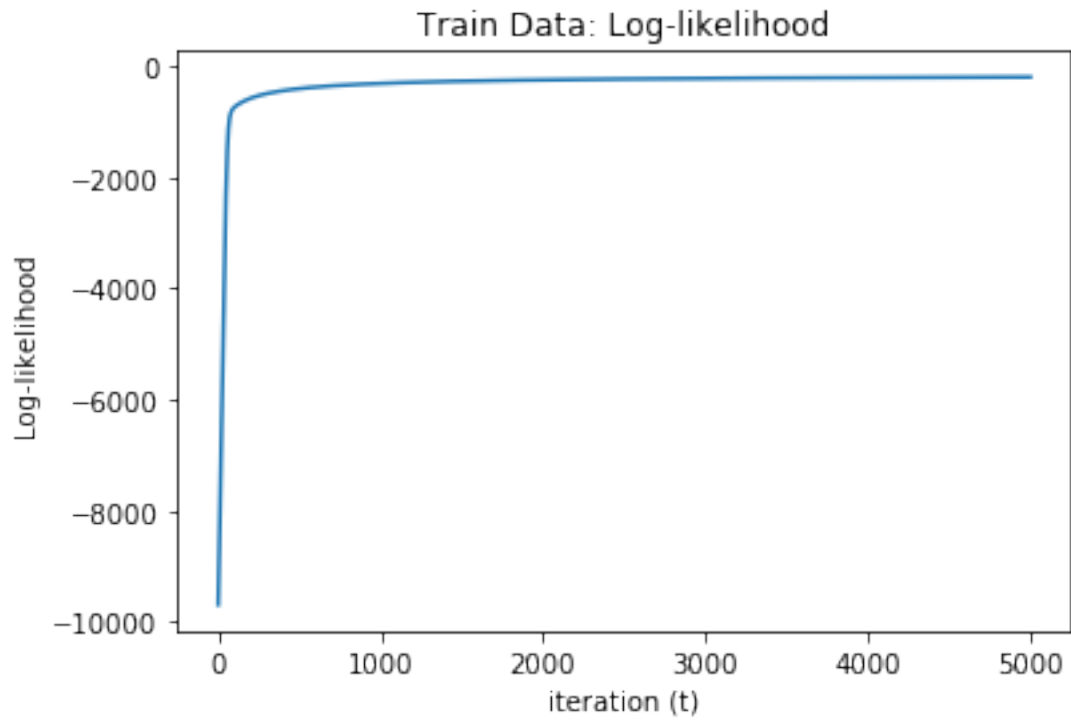
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iteration 400  
err=0.117857  
iteration 500  
err=0.109286  
iteration 600  
err=0.102857  
iteration 700  
err=0.097143  
iteration 800  
err=0.091429  
iteration 900  
err=0.087143  
iteration 1000  
err=0.082857  
iteration 1100  
err=0.080714  
iteration 1200  
err=0.077143  
iteration 1300  
err=0.072143  
iteration 1400  
err=0.072143  
iteration 1500  
err=0.072143  
iteration 1600  
err=0.069286  
iteration 1700  
err=0.067143  
iteration 1800  
err=0.067143  
iteration 1900  
err=0.067143  
iteration 2000  
err=0.067143  
iteration 2100  
err=0.065714  
iteration 2200  
err=0.066429  
iteration 2300  
err=0.065000  
iteration 2400  
err=0.063571  
iteration 2500  
err=0.062143  
iteration 2600  
err=0.062857  
iteration 2700  
err=0.062857

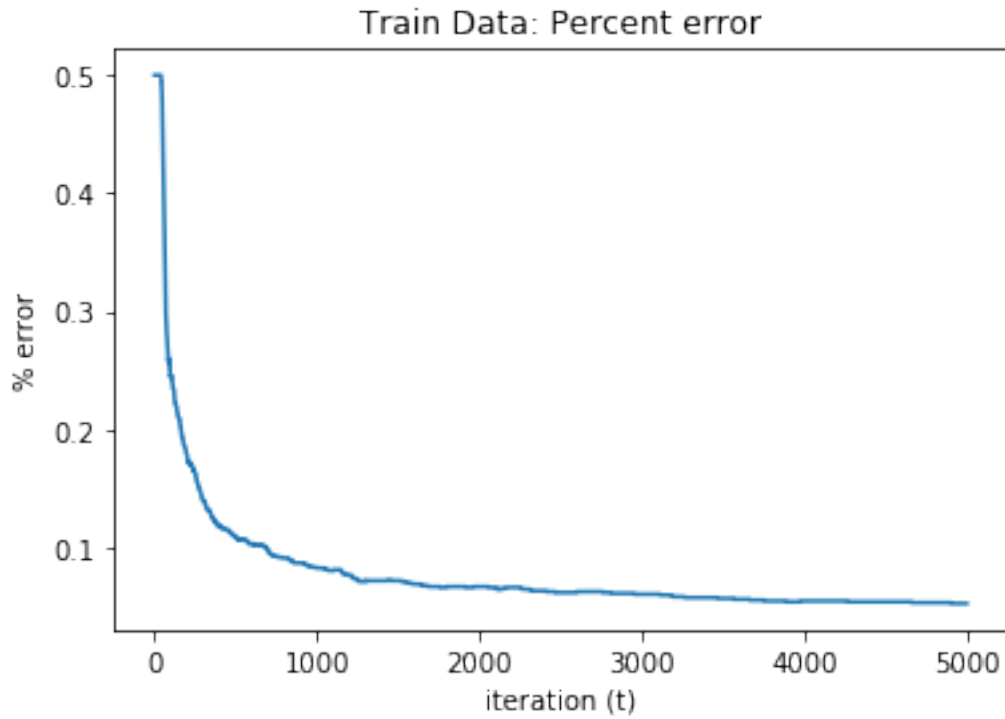
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iteration 2800
err=0.062143
iteration 2900
err=0.061429
iteration 3000
err=0.060714
iteration 3100
err=0.060714
iteration 3200
err=0.058571
iteration 3300
err=0.057857
iteration 3400
err=0.057857
iteration 3500
err=0.057143
iteration 3600
err=0.056429
iteration 3700
err=0.055714
iteration 3800
err=0.055000
iteration 3900
err=0.054286
iteration 4000
err=0.055000
iteration 4100
err=0.055000
iteration 4200
err=0.055000
iteration 4300
err=0.054286
iteration 4400
err=0.054286
iteration 4500
err=0.054286
iteration 4600
err=0.054286
iteration 4700
err=0.053571
iteration 4800
err=0.053571
iteration 4900
err=0.053571
```

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In [48]: plt.plot(Lw_train)
         plt.title("Train Data: Log-likelihood")
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plt.xlabel("iteration (t)")
plt.ylabel("Log-likelihood")
plt.savefig('log_likelihood.png')
```



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In [53]: plt.plot(err_train)
plt.title("Train Data: Percent error")
plt.xlabel("iteration (t)")
plt.ylabel("% error")
plt.savefig('error.png')
```



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In [82]: # '\n'.join([''.join(str(w_train[i:i+8])) for i in range(0,len(w_train),8)])
with open('trained_weight_vector.txt', 'w') as f:
    count = 0
    for i in range(len(w_train)):
        f.write(str(w_train[i]) + ' ')
        count += 1
        if count%8 == 0 and count>0:
            count = 0
            f.write('\n')
```

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In [84]: test_err = predict(test, test_labs, w_train)
test_err
```

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Out[84]: 0.05375
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