

CSCI-4100 Assignment 7  
Yichuan Wang  
RIN:661414395

1. (500) Classifying Handwritten Digits: 1 vs. 5

Linear regression followed by pocket algorithm improvement is used in this problem.

Math definition for feature 1 (normalized y-axis asymmetry):

$$\frac{1}{256} \sum_{i=0}^{15} \sum_{j=0}^7 |pix[i][j] - pix[i][15-j]|$$

Math definition for feature 2 (normalized x-axis asymmetry):

$$\frac{1}{256} \sum_{i=0}^7 \sum_{j=0}^{15} |pix[i][j] - pix[15-i][j]|$$

Normalization is required since original feature values are too large and will make linear regression result too small to play a role.

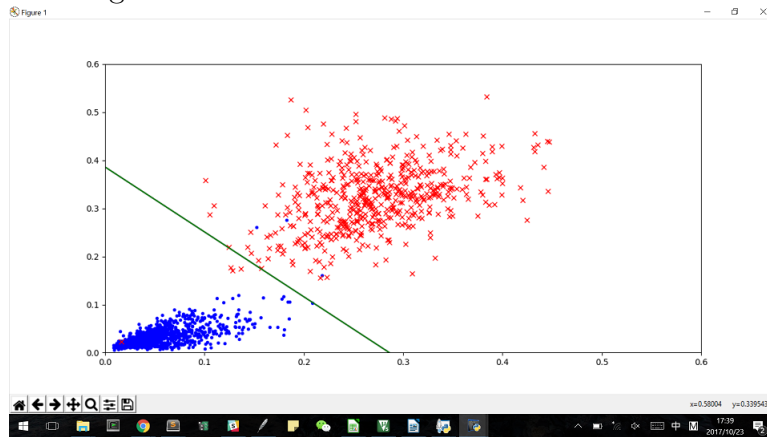
For all the plots below, feature 1 is horizontal axis, and feature 2 is vertical axis.

(a)

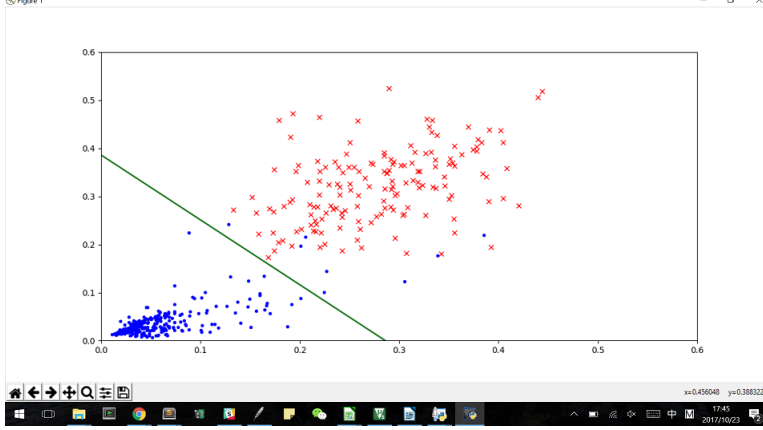
The training output  $[g_0...g_d]$  is

$$[1.2215, -4.2678, -3.1657]$$

Training data result:



Test data result:



(b)

Error Calculation: The error calculation is done in the software, the result is given here:

$$E_{in} = 0.0045$$

$$E_{test} = 0.0189$$

(c)

Error bound calculation:

$$E_{out} \leq E_{in} + \sqrt{\frac{8}{1561} \ln\left(\frac{4 \times ((1561 \times 2)^3 + 1)}{0.05}\right)} = 0.0045 + 0.3823 = 0.3868$$

$$E_{out} \leq E_{test} + \sqrt{\frac{1}{2 \times 424} \ln\left(\frac{1 \times 2}{0.05}\right)} = 0.0189 + 0.0660 = 0.0849$$

The bound based on  $E_{test}$  is a better bound.

(d)

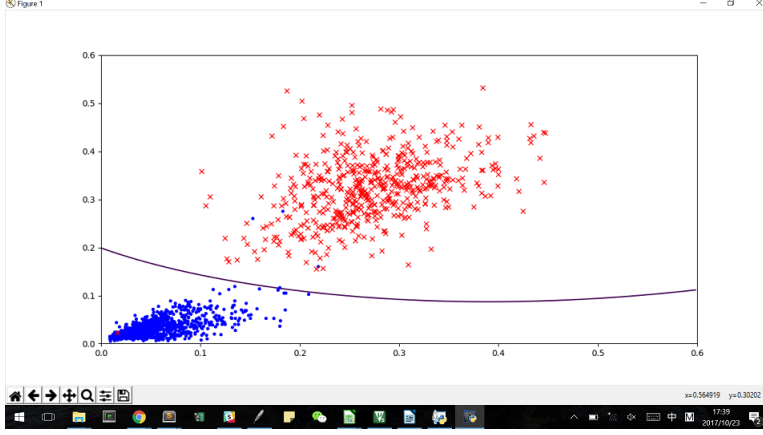
The 3rd-order transform is performed as following:

$$[x_0, x_1, x_2] \longrightarrow [x_0, x_1, x_2, x_1^2, x_2^2, x_1x_2, x_1^3, x_2^3, x_1^2x_2, x_1x_2^2]$$

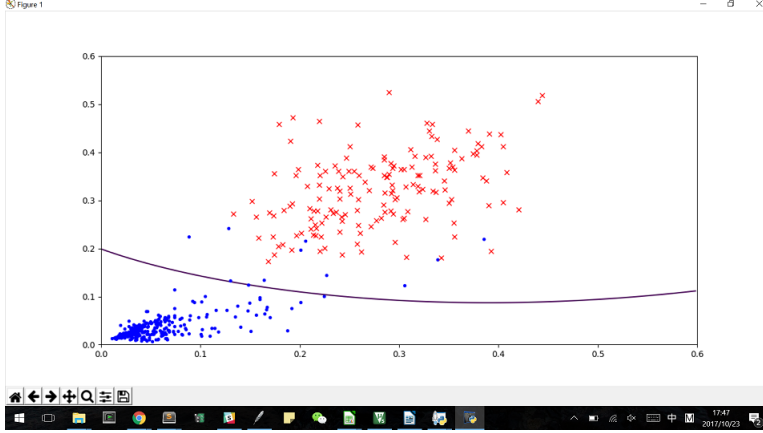
The training output  $[g_0 \dots g_d]$  is

$$[1.0567, 2.8568, -4.4757, -10.6157, -4.8360, -36.8246, 18.9741, 3.1472, -0.3295, 83.0706]$$

Training data result:  $E_{in} = 0.0026$



Test data result:  $E_{test} = 0.0213$



Error bound calculation:

$$E_{out} \leq E_{in} + \sqrt{\frac{8}{1561} \ln\left(\frac{4 \times ((1561 \times 2)^{10} + 1)}{0.05}\right)} = 0.0026 + 0.6594 = 0.6620$$

$$E_{out} \leq E_{test} + \sqrt{\frac{1}{2 \times 424} \ln\left(\frac{1 \times 2}{0.05}\right)} = 0.0213 + 0.0660 = 0.0873$$

The bound based on  $E_{test}$  is a better bound.

(e) I would like to deliver the 2D-perceptron model instead of the 3rd-order-transform one. First of all, the 2D model has a better error bar; data is

nearly separable in its original form. Second, the 3rd-order-transform model is much more complicated than the 2D model ( $d_{vc} = 3$  vs  $d_{vc} = 10$ ), and a over complicated model is likely to fit more noise.

## 2. (200) Gradient Descent on a "Simple" Function

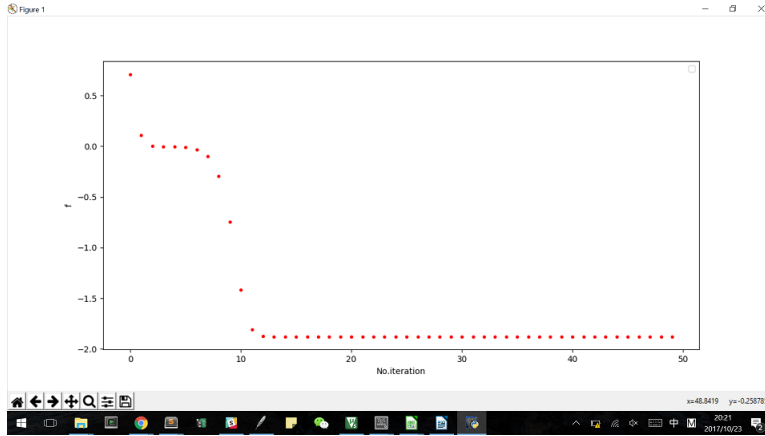
(a) Original Function:  $f(x) = x^2 + 2y^2 + 2\sin(2\pi x)\sin(2\pi y)$

$$\frac{df(x)}{dx} = 2x + 4\pi\cos(2\pi x)\sin(2\pi y)$$

$$\frac{df(x)}{dy} = 4y + 4\pi\sin(2\pi x)\cos(2\pi y)$$

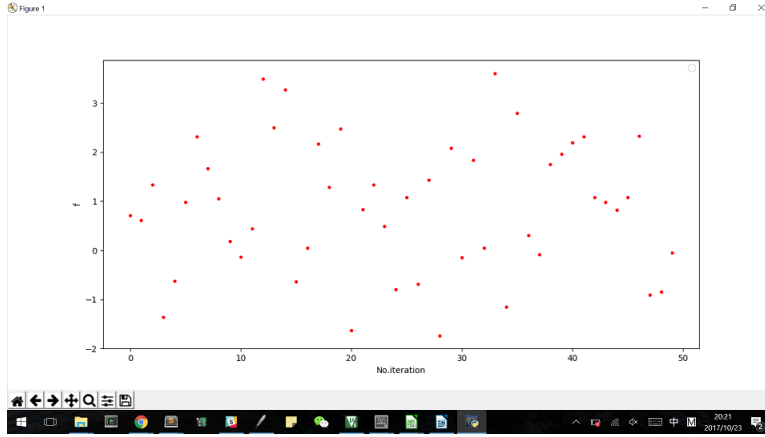
Gradient Decent result:

$\eta = 0.01$



$\eta = 0.1$

With  $\eta = 0.1$  the algorithm results in serious bouncing and cannot find the local minimum.



(b) Table:

(x,y,n)	x_final	y_final	minimum
(0.1, 0.1, 0.01)	0.243804969365	-0.237925821486	-1.87668723808
(0.1, 0.1, 0.1)	0.348357019225	-0.069582260188	-0.0446299777336
(1, 1, 0.01)	1.21807030131	0.712811950602	0.0851684974051
(1, 1, 0.1)	0.198459835146	0.457823614804	0.489676970193
(-0.5, -0.5, 0.01)	-0.731377460414	-0.237855362902	-1.38905623599
(-0.5, -0.5, 0.1)	0.591800710459	-0.125032321141	-0.723177059966
(-1, -1, 0.01)	-1.21807030131	-0.712811950602	0.0851684974051
(-1, -1, 0.1)	-0.198459835146	-0.457823614804	0.489676970193

It is observed that different starting location gives different local minimum; it is hard to find a global minimum in such condition.

### 3. (300) Problem 3.16 in LFD

(a)

$$\text{cost}(\text{accept}) = P[\text{correct}] * 0 + P[\text{intruder}] * C_a = (1 - g(x)) \times C_a$$

$$\text{cost}(\text{reject}) = P[\text{correct}] * C_r + P[\text{intruder}] * 0 = g(x) \times C_r$$

(b) It's the trade off between accept and reject. When  $\text{cost}(\text{accept}) \leq \text{cost}(\text{reject})$  we will accept the person:

$$(1 - g(x)) \times C_a \leq g(x) \times C_r$$

$$C_a - g(x) \times C_a \leq g(x) \times C_r$$

$$C_a \leq g(x) \times (C_r + C_a)$$

$$g(x) \geq \frac{C_a}{C_r + C_a}$$

$$k = \frac{C_a}{C_r + C_a}$$

(c)

Supermarket:  $C_a = 1$   $C_r = 10$

$$k = \frac{1}{10 + 1} = 0.0909$$

CIA:  $C_a = 1000$   $C_r = 1$

$$k = \frac{1000}{1000 + 1} = 0.9990$$

The intuition is that the supermarket has a high cost of false reject, so it will accept the person at a relatively low threshold. CIA has a high cost of false accept, so they accept the person only when they are super sure that it's the correct person, and thus has a high threshold.