## THE HONG KONG UNIVERSITY OF SCIENCE & TECHNOLOGY CSIT6000G/MSBD5012: Machine Learning Homework 1

Assigned: 13/09/2019 (CSIT), 14/09/2019 (MSBD), Due Date: 04/10/2019(CSIT), 05/10/2019 (MDBD),

To submit your work, hand it to the instructor on the due date.

**Question 1** Consider carrying out linear regression on the following dataset. Manually compute the ordinary least squares solution.

$x_1$	0	0	1	1	1
$x_2$	1	1	1	0	0
y	0	1	2	3	4

**Question 2** The following figures show linear regression results on a dataset of only three data points (marked blue).



The results were obtained using following regularization schemes:

- 1.  $\frac{1}{3}\sum_{i=1}^{3}(y_i w_0 w_1x_i)^2 + \lambda w_1^2$  where  $\lambda = 1$ .
- 2.  $\frac{1}{3}\sum_{i=1}^{3}(y_i w_0 w_1x_i)^2 + \lambda w_1^2$  where  $\lambda = 10$ .
- 3.  $\frac{1}{3}\sum_{i=1}^{3}(y_i w_0 w_1x_i)^2 + \lambda(w_0^2 + w_1^2)$  where  $\lambda = 1$ .
- 4.  $\frac{1}{3}\sum_{i=1}^{3}(y_i w_0 w_1x_i)^2 + \lambda(w_0^2 + w_1^2)$  where  $\lambda = 10$ .

Match the regularization schemes with the regress results. Briefly explain your answers.

**Question 3** Consider applying logistic regression to the following dataset:

$x_1$	0	0	1	1
$x_2$	0	1	0	1
y	0	0	0	1

The target is to learn a model of the form  $p(y = 1 | \mathbf{x}, \mathbf{w}) = \sigma(w_0 + w_1 x_1 + w_2 x_2)$ .

Suppose  $w_0 = -2$ ,  $w_1 = 1$  and  $w_2 = 1$  initially and  $\alpha = 0.1$ . Manually run the batch gradient descent algorithm for one iteration. Give the weights and training error after the iteration.

**Question 4** Consider applying logistic regression to the following dataset:

$x_1$	0	0	1	1
$x_2$	0	1	0	1
y	1	0	0	1

1. If we use raw feature  $x_1$  and  $x_2$ , the model is

$$p(y = 1 | \mathbf{x}, \mathbf{w}) = \sigma(w_0 + w_1 x_1 + w_2 x_2).$$

What is the minimum achievable training error in this case? Give weights that achieve the minimum error.

2. Next consider using an additional feature  $x_1x_2$  in addition to the raw feature  $x_1$  and  $x_2$ . The model now is

$$p(y = 1 | \mathbf{x}, \mathbf{w}) = \sigma(w_0 + w_1 x_1 + w_2 x_2 + w_3 x_1 x_2).$$

What is the minimum achievable training error in this case? Give weights that achieve the minimum error.

**Question 5** Consider the gradient vector in logistic regression  $\nabla_{\mathbf{w}} NNL(\mathbf{w}) = \left(\frac{\partial NNL(\mathbf{w})}{\partial w_0}, \frac{\partial NNL(\mathbf{w})}{\partial w_1}, \dots, \frac{\partial NNL(\mathbf{w})}{\partial w_D}\right)$  where

$$\frac{\partial NNL(\mathbf{w})}{\partial w_i} = -\sum_{i=1}^N [y_i - \sigma(z_i)] x_{i,j}.$$

Suppose the feature  $x_1$  is binary and, in the training set, it takes value 1 only in a small number of training examples with class label 1 (i.e., y = 1), and it takes value 0 in all training examples with class label 0 (i.e., y = 0). What will happen to the weight  $w_1$  if we update it repeatedly using the following rule:

$$w_1 \leftarrow w_1 + \alpha \sum_{i=1}^{N} [y_i - \sigma(\mathbf{w}^\top \mathbf{x}_i)] x_{i,1}$$

What if we use the following update rule instead:

$$w_1 \leftarrow w_1 + \alpha [-\lambda w_1 + \sum_{i=1}^N [y_i - \sigma(\mathbf{w}^\top \mathbf{x}_i)] x_{i,1}],$$

where  $\lambda$  is the regularization constant?