

CS188, Winter 2017
Problem Set 2:
Due 2/9/2017

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1 Problem 1

- (a) $w_1 = 1, w_2 = 1, b = -1$
- (b) XOR's feature values are not linearly separable so it cannot be represented by a perceptron.

2 Problem 2

(a)

$$h_\theta = \frac{1}{1 + e^{-\theta x_n}} \quad (1)$$

$$\log(h_\theta(x_n)) = -\log(1 + e^{-\theta x_n}) \quad (2)$$

$$\log(1 - h_\theta(x_n)) = -\theta x_n - \log(1 + e^{-\theta x_n}) \quad (3)$$

$$J(\theta) = \sum_{n=1}^N y_n \log(1 + e^{-\theta x_n}) + (1 - y_n)(\theta x_n + \log(1 + e^{-\theta x_n})) \quad (4)$$

$$= \sum_{n=1}^N -y_n \theta x_n + \theta x_n + \log(1 + e^{-\theta x_n}) \quad (5)$$

$$= \sum_{n=1}^N -y_n \theta x_n + \log(1 + e^{\theta x_n}) \quad (6)$$

$$\frac{\partial J(\theta)}{\partial \theta} = \sum_{n=1}^N -x_n y_n + \frac{x_n e^{\theta x_n}}{e^{\theta x_n} + 1} \quad (7)$$

$$\frac{\partial J(\theta)}{\partial \theta} = \sum_{n=1}^N -x_n y_n + x_n h_\theta(x_n) \quad (8)$$

(b)

$$\frac{\partial^2 J(\theta)}{\partial \theta_j \partial \theta_k} = \sum_{n=1}^N x_{n_j} h'_\theta(x_n) \quad (9)$$

because

$$h'_\theta(x_n) = h_\theta(x_n)(1 - h_\theta(x_n)), \quad (10)$$

$$\frac{\partial^2 J(\theta)}{\partial \theta_j \partial \theta_k} = \sum_{n=1}^N x_{n_j} h_\theta(x_n)(1 - h_\theta(x_n)) x_n^T \quad (11)$$

(c) Because H is written as

$$H = \sum_{n=1}^N x_{n_j} h_\theta(x_n)(1 - h_\theta(x_n)) x_n^T, \quad (12)$$

and we know that the function h, representing the sigmoid function, must be between 0 and 1 hence making

$$h_\theta(x_n)(1 - h_\theta(x_n)) \quad (13)$$

positive. For J to be convex, H must be positive semi-definite. Therefore we can exclude Equation 13 from Equation 12, only paying attention to x_n .

$$z^T x_n x_n^T z = (x_n^T z)^T (x_n^T z) = (x_n^T z) \cdot (x_n^T z) \quad (14)$$

The dot product is positive which means that the Hessian matrix is positive semi-definite thus J is convex.

3 Problem 3

(a)

$$\frac{\partial J(\theta)}{\partial \theta_0} = 2 \sum_{n=1}^N w_n (\theta_0 + \theta_1 x_{n,1} - y_n) \quad (15)$$

$$\frac{\partial J(\theta)}{\partial \theta_1} = 2 \sum_{n=1}^N w_n (\theta_0 + \theta_1 x_{n,1} - y_n) x_{n,1} \quad (16)$$

(b) A system of equations was made from part (a) as follows:

$$\sum_{n=1}^N w_n y_n = \theta_0 \sum_{n=1}^N w_n + \theta_1 \sum_{n=1}^N w_n x_{n,1} \quad (17)$$

$$\sum_{n=1}^N w_n x_{n,1} y_n = \theta_0 \sum_{n=1}^N w_n x_{n,1} + \theta_1 \sum_{n=1}^N w_n x_{n,1}^2 \quad (18)$$

Solving for theta 0 and 1:

$$\theta_0 = \frac{\sum_{n=1}^N w_n y_n - \theta_1 \sum_{n=1}^N w_n x_{n,1}}{\sum_{n=1}^N w_n} \quad (19)$$

Plugging it into Equation 18, I can solve for theta 1:

$$\theta_1 = \frac{\sum_{n=1}^N w_n \sum_{n=1}^N w_n x_{n,1} y_n - \sum_{n=1}^N w_n y_n \sum_{n=1}^N w_n x_{n,1}}{\sum_{n=1}^N w_n \sum_{n=1}^N w_n x_{n,1}^2 - (\sum_{n=1}^N w_n x_{n,1})^2} \quad (20)$$

Plugging Equation 20 back into Equation 19, we can solve for theta 0 as well:

$$\theta_0 = \frac{\sum_{n=1}^N w_n y_n}{\sum_{n=1}^N w_n} - \frac{\sum_{n=1}^N w_n x_{n,1}}{\sum_{n=1}^N w_n} \cdot \theta_1 \quad (21)$$