

CSCI567 2017 Homework Assignment 1 Solution

Algorithmic component

1 Linear Regression

Question 1.1 $X^T X$ is not invertible

$$r(X^T X) \geq D + 1,$$

where $r(M)$ is the rank of matrix M .

Question 1.2 Bias solution

$$b^* = \arg \min_b \|\mathbf{y} - b\mathbf{1}_N - X^T \mathbf{w}\|^2 \quad \text{Residual sum of squares} \quad (1)$$

$$\mathbf{1}_N^T (\mathbf{y} - b^* \mathbf{1}_N - X^T \mathbf{w}) = 0 \quad \text{Taking derivatives w.r.t } b \quad (2)$$

$$b^* = \frac{1}{N} (\mathbf{1}_N^T \mathbf{y} - \mathbf{1}_N^T X^T \mathbf{w}) \quad \text{solve for } b^* \quad (3)$$

$$= \frac{1}{N} \mathbf{1}_N^T \mathbf{y} \quad \frac{1}{N} \sum_n x_{nd} = 0 \Leftrightarrow \mathbf{1}_N^T X^T = \mathbf{0} \quad (4)$$

If the feature values are zero on average, the bias b^* is the average response of training samples.

2 Logistic Regression

Question 2.1 Bias solution

$$b^* = \min_b - \sum_n \{y_n \log \sigma(b) + (1 - y_n) \log[1 - \sigma(b)]\} \quad \text{cross entropy objective} \quad (5)$$

$$\sum_n y_n (1 - \sigma(b^*)) - (1 - y_n) \sigma(b^*) = 0 \quad \text{Taking derivatives w.r.t } b \quad (6)$$

$$\sigma(b^*) = \frac{\sum_n y_n}{N} \quad \text{solve for } b^* \quad (7)$$

$$b^* = \log\left(\frac{\sum_n y_n}{\sum_n (1 - y_n)}\right) \quad (8)$$

$\sigma(b^*)$ is the optimal logistic regression classifier, also the probability that a test sample is labeled as 1, which is the fraction of label 1 samples in the training data.