

## IE 7374: Machine Learning

Prove that if negative-log likelihood for Logistic regression is Convex.

### Answer to Question

$$X = \{x_1, \dots, x_i, \dots, x_n\}_{n \times d}^T, x_i \in \mathbb{R}^d$$
$$p(x_i|\theta) = \frac{1}{1 + \exp(-\theta x_i)}$$

$$\begin{aligned} P &= -\log[p(X|\theta)] \\ &= -\log\left[\sum_{i=1}^n p(x_i|\theta)\right] \\ &= \sum_{i=1}^n \log[1 + \exp(-\theta x_i)] \end{aligned}$$

Then, we calculate the second derivative of P with respect to  $\theta$ , to see if it is greater or equal than 0. If so, we prove it is a convex function.

$$\begin{aligned} \frac{\partial P}{\partial \theta} &= \sum_{i=1}^n \frac{1}{1 + \exp(-\theta x_i)} \cdot \exp(-\theta x_i) \cdot (-x_i) \\ \frac{\partial^2 P}{\partial \theta^2} &= \sum_{i=1}^n [-(1 + \exp(-\theta x_i))^{-2} \cdot \exp(-\theta x_i) \cdot (-x_i) \cdot \exp(-\theta x_i) \cdot (-x_i) + \\ &\quad (1 + \exp(-\theta x_i))^{-1} \cdot \exp(-\theta x_i) \cdot (-x_i) \cdot (-x_i)] \\ &= \sum_{i=1}^n [(1 + \exp(-\theta x_i))^{-2} \cdot \exp(-2\theta x_i) \cdot x_i^2] [(1 + \exp(-\theta x_i)) \cdot \exp(\theta x_i) - 1] \\ &= \sum_{i=1}^n [(1 + \exp(-\theta x_i))^{-2} \cdot \exp(-2\theta x_i) \cdot x_i^2] \cdot \exp(\theta x_i) \\ &\geq 0 \end{aligned}$$

Negative-log likelihood for logistic regression is a Convex.