

Assignment 1

Written assignment

1. Consider stochastic gradient descent method to learn the house price model

$$h(x_1, x_2) = \sigma(b + w_1x_1 + w_2x_2),$$

where σ is the sigmoid function.

Given one single data point $(x_1, x_2, y) = (1, 2, 3)$, and assuming that the current parameter is $\theta^0 = (b, w_1, w_2) = (4, 5, 6)$, evaluate θ^1 .

Just write the expression and substitute the numbers; no need to simplify or evaluate.

2. (a) Find the expression of $\frac{d^k}{dx^k} \sigma$ in terms of $\sigma(x)$ for $k = 1, \dots, 3$ where σ is the sigmoid function.
- (b) Find the relation between sigmoid function and hyperbolic function.
3. There are unanswered questions during the lecture, and there are likely more questions we haven't covered. Take a moment to think about them and write them down here.

(1) Recall: gradient decent algorithm.

$$\theta^{n+1} = \theta^n - \alpha \nabla \text{Loss}, \quad \alpha \text{ is learning rate and } \alpha > 0.$$

For MSE loss

$$\Rightarrow \theta^1 = \theta^0 + 2\alpha \left[\frac{1}{1} \sum_{i=1}^1 (3 - h(1, 2)) \nabla_{\theta} h \right] \#$$

(2) - a.

$$(i) \sigma'(x) = \frac{d}{dx} (1 + e^x)^{-1}$$

$$= (1 + e^x)^{-2} \cdot e^x$$

$$= (1 + e^x) \cdot \left(\frac{e^x}{1 + e^x} \right)$$

$$= \sigma(x) \cdot (1 - \sigma(x)) \quad \#$$

$$(ii) \sigma''(x) = \sigma'(x) \cdot (1 - \sigma(x)) + \sigma(x) (1 - \sigma(x))'$$

$$= \sigma(x) (1 - \sigma(x))^2 + \sigma(x) \cdot (-1) \cdot \sigma'(x)$$

$$= \sigma(x) (1 - \sigma(x))^2 - \sigma(x) \cdot \sigma(x) \cdot (1 - \sigma(x))$$

$$= \sigma(x) (1 - \sigma(x)) (1 - \sigma(x) - \sigma(x))$$

$$= \sigma(x) \cdot (1 - \sigma(x)) \cdot (1 - 2\sigma(x)) \quad \#$$

$$(iii) \sigma^{(3)}(x) = \sigma'(x) (1 - \sigma(x)) (1 - 2\sigma(x))$$

$$+ \sigma(x) (1 - \sigma(x))' (1 - 2\sigma(x))$$

$$+ \sigma(x) (1 - \sigma(x)) (1 - 2\sigma(x))'$$

$$= \sigma(x) (1 - \sigma(x))^2 (1 - 2\sigma(x))$$

$$+ \sigma(x) (-1) \sigma(x) (1 - \sigma(x)) (1 - 2\sigma(x))$$

$$+ \sigma(x) \cdot (1 - \sigma(x)) \cdot (-2) \cdot \sigma(x) (1 - \sigma(x))$$

$$= \sigma(x) (1 - \sigma(x)) (1 - 6\sigma(x) + 6\sigma^2(x)) \quad \#$$

(2) - b.

$$\tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}} = \frac{1 - e^{-2x}}{1 + e^{-2x}} = \frac{1}{1 + e^{-2x}} (1 - e^{-2x}).$$

$$\Rightarrow \frac{1}{1 + e^{-2x}} (1 - e^{-2x}) = 2\sigma(2x) - 1 \neq$$

13)

Q: Assume $h(x_1, x_2) = b + w_1 x_1 + w_2 x_2 = [1, x_1, x_2] \begin{bmatrix} b \\ w_1 \\ w_2 \end{bmatrix}$.

$$\begin{array}{c} \begin{bmatrix} y^1 \\ \vdots \\ y^N \end{bmatrix} \\ Y \end{array} = \begin{array}{c} \begin{bmatrix} 1 & x_1^1 & x_2^1 \\ \vdots & \vdots & \vdots \\ 1 & x_1^N & x_2^N \end{bmatrix} \\ \underbrace{\hspace{1.5cm}}_X \end{array} \begin{array}{c} \begin{bmatrix} b \\ w_1 \\ w_2 \end{bmatrix} \\ \theta \end{array}.$$

$$\min \|Y - X\theta\|_2^2 \Rightarrow \theta^* = (X^T X)^{-1} X^T Y \quad ?$$

A: Goal: 找到 1 個向量 $X\theta - Y$ 垂直於 $\text{col}(X)$ 的平面上.

$$\therefore \min \|Y - X\theta\|_2^2$$

$$\Rightarrow Y - X\theta \in \text{col}(X)^\perp = N(X^T).$$

$$\Rightarrow X^T (X\theta - Y) = 0.$$

Suppose X is full rank $\Rightarrow \text{rank}(X) = N$

$$\Rightarrow \text{rank}(X^T X) = \text{rank}(X) \Rightarrow \text{rank}(X^T X) = N.$$

故 $X^T X$ is full rank, $\det(X^T X) \neq 0$,

$\Rightarrow X^T X$ is invertible.

Then, $\theta = (X^T X)^{-1} X^T Y$ #

Q: 上課中提到選取 learning rate 時,

有說可以在每一步選取一大一小的 α .

然後可以找到 $\arg\min_{\theta} \text{Loss}(\theta)$, why?

A: 應該是要保有 ① 加速逼近最佳點.

和 ② 保持穩定, 避免跳過最佳點.

⇒ 使用這種方法在取大的 learning rate 時,
要讓它在穩定範圍內,
讓小的 learning rate 將它拉回來.

問
chatGPT