HW1 2025/09/07

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 .

Solution:

Assume the loss function is $L(\theta) = \frac{1}{n} \sum_{i=1}^n ||y-h(x_1,x_2)||_2^2 = (y-h(x_1,x_2))^2$. Implement stochastic gradient descent method.

$$\theta^{1^T} = \theta^{0^T} - \alpha \nabla_{\theta} L$$

where α is the learning rate.

$$heta^{1^T} = egin{pmatrix} b^1 \ w_1^1 \ w_2^1 \end{pmatrix} = egin{pmatrix} 4 \ 5 \ 6 \end{pmatrix} + 2lpha(y-h(x_1,x_2))h'(x_1,x_2)$$

$$heta^{1^T} = egin{pmatrix} 4 \ 5 \ 6 \end{pmatrix} + 2lpha(3-\sigma(4+5\cdot 1+6\cdot 2))\sigma'(x_1,x_2,b) egin{pmatrix} 1 \ 1 \ 2 \end{pmatrix}$$

Since the derivative of sigmoid function is $\sigma'(z)=-(1+e^{-z})^{-2}e^{-z}\cdot(-1)=\frac{e^{-z}}{(1+e^{-z})^2}=\sigma(z)(1-\sigma(z))$, then

$$heta^{1T} = egin{pmatrix} 4 \ 5 \ 6 \end{pmatrix} + 2lpha(3-\sigma(21))\sigma(21)(1-\sigma(21)) egin{pmatrix} 1 \ 1 \ 2 \end{pmatrix}$$

2.(a) Find the experssion of $rac{d^k}{dx^k}\sigma$ in terms of $\sigma(x)$ for $k=1,\ldots,3$ where σ is the sigmoid function.

Solution:

In problem 1, I have showed the derivative of sigmoid function.

$$\frac{d}{dx}\sigma(x) = \sigma(x)(1 - \sigma(x))$$

$$\frac{d^2}{dx^2}\sigma(x) = \sigma'(x)(1 - \sigma(x)) - \sigma'(x)\sigma(x)$$
$$= \sigma'(x) - 2\sigma(x)\sigma'(x)$$
$$= (1 - 2\sigma(x))\sigma'(x)$$
$$= (1 - 2\sigma(x))\sigma(x)(1 - \sigma(x))$$

$$\frac{d^3}{dx^3}\sigma(x) = \sigma'(x)(1 - \sigma(x))(1 - 2\sigma(x)) + \sigma(x)(-\sigma'(x))(1 - 2\sigma(x)) + \sigma(x)(1 - \sigma(x))(-2\sigma'(x))$$

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

(b) Find the relation between sigmoid function and hyperbolic function.

Solution:

Let the logistic sigmoid be

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

Step 1 — Solve for the exponentials in terms of $\sigma(x)$:

$$e^{-x} = rac{1}{\sigma(x)} - 1 = rac{1 - \sigma(x)}{\sigma(x)}, \qquad e^x = rac{1}{e^{-x}} = rac{\sigma(x)}{1 - \sigma(x)}.$$

Step 2 — Start from the definition of tanh:

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

Step 3 — Substitute the expressions from Step 1:

$$anh(x) = rac{rac{\sigma}{1-\sigma} - rac{1-\sigma}{\sigma}}{rac{\sigma}{1-\sigma} + rac{1-\sigma}{\sigma}} = rac{\sigma^2 - (1-\sigma)^2}{\sigma^2 + (1-\sigma)^2}.$$

Step 4 — Simplify the numerator and denominator:

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

$$\sigma^2 + (1 - \sigma)^2 = \sigma^2 + (1 - 2\sigma + \sigma^2) = 2\sigma^2 - 2\sigma + 1 = \sigma^2 + (\sigma - 1)^2.$$

Result:

$$anh(x) = rac{2\sigma(x) - 1}{\sigma^2(x) + ig(\sigma(x) - 1ig)^2}$$

3.在課堂上有提到過 optimizer 的策略的問題,教授你有提到有論文發表了新的optimizer,它的想法是先走一大步,再走一小步,然後再走一大步,接著一小步,如此的循環反覆更新參數,最後就會到最佳解,那這個還蠻有趣的,因為目前optimizer 大家比較常用的是Adam,想知道說當把optimizer的更新過程給視覺化後,跟Adam的差別是如何?