

$$\frac{dL}{d\bar{z}^{[2]}} = \frac{dL}{da^{[2]}} \cdot \frac{da^{[2]}}{d\bar{z}^{[2]}} = a^{[2]} - y$$

$$\frac{dL}{dW^{[2]}} = \frac{dL}{d\bar{z}^{[2]}} \cdot \frac{d\bar{z}^{[2]}}{dW^{[2]}} = (a^{[2]} - y) \cdot a^{[1]T}$$

$$\frac{dL}{db} = \frac{dL}{d\bar{z}^{[2]}} \cdot 1 = d\bar{z}^{[2]}$$

$$\begin{aligned} \frac{dL}{d\bar{z}^{[1]}} &= \frac{dL}{d\bar{z}^{[2]}} \cdot \frac{d\bar{z}^{[2]}}{d\bar{z}^{[1]}} \\ &= \frac{dL}{d\bar{z}^{[2]}} \cdot \frac{d\bar{z}^{[2]}}{da^{[2]}} \cdot \frac{da^{[2]}}{d\bar{z}^{[1]}} \\ &= \frac{dL}{d\bar{z}^{[2]}} \cdot W^{[2]} \end{aligned}$$

$$= W^{[2]T} d\bar{z}^{[2]}, \quad g^{[2]'}(\bar{z}^{[1]})$$

$$dW^{[2]} = d\bar{z}^{[2]} x^T$$

$$db^{[2]} = d\bar{z}^{[2]}$$

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为什么无偏估计的方差是 $\frac{\sigma^2}{n-1}$?

$$\sigma^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

$$E(XY) = EX \cdot EY$$

$$\text{Var}(\bar{X}) = \frac{1}{n} \text{Var}(X) \quad ?$$

$$\sigma(\bar{X})^2 = \frac{1}{n} \sigma(X)^2$$

二阶中心距:

$$\begin{aligned} \text{Var}(X) &= E(X^2) - (E(X))^2 \\ E(X^2) &= \text{Var}(X) + (E(X))^2 = \sigma^2 + \mu^2 \end{aligned}$$

$$E(\bar{X}^2) = \text{Var}(\bar{X}) + (E(\bar{X}))^2 = \frac{1}{n} \sigma^2 + \mu^2$$

无偏估计: 对变量 θ 的估计是 $\hat{\theta}$, 如果 $E(\hat{\theta}) = E(\theta)$, 则称 $\hat{\theta}$ 为 θ 的无偏估计.

$$E\left(\sum_{i=1}^n (x_i - \bar{x})^2\right)$$

$$= \sum_{i=1}^n E(x_i^2) - n \cdot E(\bar{x}^2)$$

$$= \sum_{i=1}^n (\sigma^2 + \mu^2) - n \left(\frac{1}{n} \sigma^2 + \mu^2 \right)$$

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