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3C02

①

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$x_1, x_2, \dots, x_n \rightarrow$ sample size of n

$$L(x_1, x_2, \dots, x_n) = f(x_1) \cdot f(x_2) \cdot \dots \cdot f(x_n)$$

$$\Rightarrow \left(\frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x_1-\mu)^2}{2\sigma^2}} \right) \cdot \left(\frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x_2-\mu)^2}{2\sigma^2}} \right) \cdot \dots$$

$$\ln(L) = -\frac{n}{2} \ln(2\pi\sigma^2) + \sum_{i=1}^n \left(-\frac{(x_i - \mu)^2}{2\sigma^2} \right) \quad \text{--- ①}$$

taking partial derivative w.r.t μ

$$\begin{aligned} \frac{\partial \ln(L)}{\partial \mu} &= 0 + \sum_{i=1}^n -\left(\frac{(x_i - \mu)}{\sigma^2} \right) \\ &= \sum_{i=1}^n (x_i - \mu) = 0 \end{aligned}$$

$$n\bar{x} - n\mu = 0$$

$$\bar{x} = \mu$$

$\mu = \bar{x} \Rightarrow$ sample mean

Taking partial derivative of eqn 1 w.r.t σ^2

$$\begin{aligned} \frac{\partial \ln(L)}{\partial \sigma^2} &= -\frac{n}{2\sigma^2} + \sum \frac{-(x_i - \mu)^2}{2(\sigma^2)^2} = 0 \\ -n + \sum \frac{-(x_i - \mu)^2}{\sigma^2} &= 0 \end{aligned}$$

$$n = \sum \frac{(x_i - \mu)^2}{\sigma^2}$$

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

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

②

Binomial distribution

$$\rightarrow {}^n C_{x_i} \theta^{x_i} (1-\theta)^{n-x_i}$$

$$L = \prod_{i=1}^n {}^n C_{x_i} \theta^{x_i} (1-\theta)^{n-x_i}$$

Take log on both sides

$$\log L = \sum_{i=1}^n \left(\log ({}^n C_{x_i}) + \log \theta^{x_i} + \log (1-\theta)^{n-x_i} \right)$$

differentiate w.r.t θ

$$\frac{d \log(L)}{d\theta} = 0$$

$$\frac{1}{\theta} \sum x_i - \frac{n}{1-\theta} + \frac{1}{1-\theta} \sum x_i = 0$$

$$\frac{1}{\theta + (1-\theta)} \sum x_i = \frac{n}{1-\theta} \Rightarrow \boxed{\theta = \frac{\sum x_i}{n}}$$