

$$Q1) \theta_1 = \mu, \theta_2 = \sigma^2$$

$$MLE = ?$$

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

$$f(x, \theta_1, \theta_2) = \frac{1}{\sqrt{2\pi\theta_2}} e^{-\frac{(x-\theta_1)^2}{2\theta_2}}$$

$$\theta_1 \in (-\infty, \infty), \theta_2 \in [0, \infty)$$

$$L(\theta_1, \theta_2) = \prod_{i=1}^n f(x_i, \theta_1, \theta_2) =$$

$$= \theta_2^{-n/2} (2\pi)^{-n/2} e^{-\frac{1}{2\theta_2} \sum_{i=1}^n (x_i - \theta_1)^2}$$

$$\begin{aligned} \log L(\theta_1, \theta_2) &= -n/2 \log \theta_2 - n/2 \log(2\pi) \\ &\quad - \frac{\sum_{i=1}^n (x_i - \theta_1)^2}{2\theta_2} \end{aligned}$$

Partial derivation both sides w.r.t
 θ_1

$$\frac{\partial \log L(\theta_1, \theta_2)}{\partial \theta_1} = \frac{-2 \sum (x_i - \theta_1)(-1)}{2\theta_2}$$

$$\sum (x_i - \theta_1) = 0$$

$$\hat{\theta}_1 = \mu = \frac{\sum x_i}{n} = \bar{x}$$

$$\frac{\partial \log L(\theta_1, \theta_2)}{\partial \theta_2} = \frac{-n}{2\theta_2} + \frac{\sum (x_i - \theta_1)^2}{2\theta_2^2}$$

$$= 0$$

$$-n\theta_2 + \sum (x_i - \theta_1)^2 = 0$$

$$\boxed{\theta_2 \neq 0}$$

$$\theta_2^{\wedge} = \sigma_2^{\wedge} = \frac{\sum (x_i - \bar{x})^2}{n}$$

$$\hat{\mu} = \frac{\sum x_i}{n}$$

$$\hat{\sigma}_2^2 = \frac{\sum (x_i - \bar{x})^2}{n}$$

$$Q2) L(\theta | x_1, x_2, \dots, x_n) =$$

$$\prod_{i=1}^n \binom{m}{x_i} \theta^{x_i} (1-\theta)^{m-x_i}$$

Taking log-likelihood

$$L(\theta | x_1, x_2, \dots, x_n) =$$

$$\sum_{i=1}^n \left[\log \binom{m}{x_i} + x_i \log \theta + (m-x_i) \log(1-\theta) \right]$$

$$\frac{\partial L}{\partial \theta} = \frac{\partial}{\partial \theta} \left(\sum_{i=1}^n \left[\log \binom{m}{x_i} + x_i \log \theta + (m-x_i) \log(1-\theta) \right] \right)$$

$$\frac{\partial L}{\partial \theta} = \sum_{i=1}^n \left[\frac{x_i}{\theta} - \frac{m-x_i}{1-\theta} \right]$$

$$\text{When } \frac{\partial L}{\partial \theta} = 0$$

$$\sum_{i=1}^n \left[\frac{x_i}{\theta} - \frac{m-x_i}{1-\theta} \right] = 0$$

$$\sum_{i=1}^n \left[\frac{x_i - \theta_m}{\theta(1-\theta)} \right] = 0$$

$$\sum_{i=1}^n \left[\frac{x_i}{\theta(1-\theta)} \right] - \sum_{i=1}^n \left[\frac{\theta_m}{\theta(1-\theta)} \right] = 0.$$

$$\sum_{i=1}^n \left[\frac{x_i}{\theta(1-\theta)} \right] = \frac{mn}{1-\theta}.$$

$$\sum_{i=1}^n x_i = mn\theta$$

$$\theta = \frac{\sum_{i=1}^n x_i}{mn}.$$