

The normal distribution

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

$$\begin{aligned} \ell(x|\mu, \sigma) &= -\frac{1}{2} \ln(2\pi\sigma^2) - \frac{1}{2} \left(\frac{x-\mu}{\sigma}\right)^2 \\ &= -\frac{1}{2} \ln(2\pi) - \ln(\sigma) - \frac{1}{2} (x-\mu)^2 \sigma^{-2} \end{aligned}$$

$$\partial_{\sigma} \ell(x|\mu, \sigma) = -\frac{1}{\sigma} + (x-\mu)^2 \sigma^{-3}$$

$$\partial_{\sigma}^2 \ell(x|\mu, \sigma) = \sigma^{-2} - 3(x-\mu)^2 \sigma^{-4}$$

$$\partial_{\mu} \partial_{\sigma} = 2(x-\mu) \sigma^{-3}$$

$$\partial_{\mu} \ell(x|\mu, \sigma) = -(x-\mu) \sigma^{-2}$$

$$\partial_{\mu}^2 \ell(x|\mu, \sigma) = \sigma^{-2}$$

$$n L_n(x|\theta) = -\frac{n}{2} (\ln 2\pi + 2\ln \sigma) - \frac{1}{2\sigma^2} \sum_i (x_i - \mu)^2$$

$$\partial_{\mu} L_n(x|\theta) = -\frac{1}{n\sigma^2} \sum_i (x_i - \mu)$$

$$\text{taking } \partial_{\mu} L_n = 0 \Rightarrow \sum_i x_i = \sum \mu = n\mu$$

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

$$\partial_{\sigma} L_n(x|\theta) = -\frac{n}{\sigma} + \frac{1}{\sigma^3} \sum_i (x_i - \mu)^2$$

$$\text{taking } \partial_{\sigma} L_n = 0 \Rightarrow n\sigma^3 = \sigma \sum_i (x_i - \mu)^2$$

$$\Rightarrow \hat{\sigma}^2 = \frac{1}{n} \sum_i (x_i - \mu)^2$$

$$I_{ij} = E_{\theta_0} \begin{pmatrix} n\sigma^{-2} & 2n(\bar{x}-\mu)\sigma^{-3} \\ 2n(\bar{x}-\mu)\sigma^{-3} & n\sigma^{-2} - 3n(\bar{x}-\mu)^2\sigma^{-4} \end{pmatrix}$$

$$I_{ij} = \begin{pmatrix} n\sigma_0^{-2} & 0 \\ 0 & n\sigma_0^{-2} \end{pmatrix}$$

Expectations & Variances

$$E(\hat{\mu}) = E(\bar{X}) = \mu$$

$$\text{Var}(\hat{\mu}) = \text{Var}(\bar{X}) = \frac{1}{n^2} \text{Var}(\sum X_i) = \frac{1}{n^2} \sum \text{Var}(X_i) = \frac{\sigma_0^2}{n}$$

$$\hat{\mu} - \mu \sim \mathcal{N}(0, \frac{1}{nI_0}) \quad \text{where } I_0 = \frac{1}{\sigma_0^2} \quad \text{by CLT}$$

where $\hat{\mu}$ is maximizer of $L_n(X|\mu, \sigma^2)$

\Rightarrow Variance of MLE for large sample size

is asymptotic to $\frac{1}{nI_0}$ (holds in general)

$$E(\hat{\sigma}^2) = E\left[\frac{1}{n} \sum (x_i - \mu)^2\right] = E\left[\frac{1}{n} \sum \left((x_i - \bar{x})^2 + 2(x_i - \bar{x})(\bar{x} - \mu) + (\bar{x} - \mu)^2\right)\right]$$

$$= E[\dots]$$

$$\text{Var}(\hat{\sigma}^2) = \text{Var}\left(\frac{1}{n} \sum (x_i - \mu)^2\right) \quad ; \quad \sum (x_i - \bar{x} + \bar{x} - \mu)^2$$

$$= \text{Var}\left[\frac{1}{n} \sum \left((x_i - \bar{x})^2 + 2(x_i - \bar{x})(\bar{x} - \mu) + (\bar{x} - \mu)^2\right)\right]$$