

Suggested Title: © Edit

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③

a) Each dataset has likelihood function

$$f_X(\sigma^2) = \prod_{i=1}^n \frac{1}{\sqrt{2\pi\sigma^2}} \cdot \exp\left\{-\frac{1}{2} \cdot \frac{x_i^2}{\sigma^2}\right\}$$

$$= (2\pi\sigma^2)^{-n/2} \cdot \exp\left\{-\frac{1}{2\sigma^2} \cdot \sum_{i=1}^n x_i^2\right\}$$

$$\ell_X(\sigma^2) = -\frac{n}{2} \log(2\pi\sigma^2) - \frac{1}{2\sigma^2} \cdot \sum_{i=1}^n x_i^2$$

$$\frac{\partial}{\partial \sigma^2} \ell_X(\sigma^2) = -\frac{n}{2} \cdot \frac{2/\pi}{2\pi\sigma^2} + \frac{1}{2(\sigma^2)^2} \cdot \sum_{i=1}^n x_i^2$$

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



Setting this expression to zero to find



Setting this expression to zero to find the MLE...

$$0 = -\frac{1}{2} \left[ \frac{n}{\sigma^2} - \frac{\sum_{i=1}^n x_i^2}{(\sigma^2)^2} \right]$$

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

$$\Rightarrow \hat{\sigma}^2 = \frac{\sum_{i=1}^n x_i^2}{n}$$

Use this to get MLE estimates  $\hat{\sigma}_1^2, \hat{\sigma}_2^2$

b) we know

$$\dot{\ell}_x(\sigma^2) = -\frac{n}{2\sigma^2} + \frac{\sum_{i=1}^n x_i^2}{2(\sigma^2)^2}$$



$$\sum_{i=1}^n x_i^2$$

Thus

$$\ddot{\ell}_x(\sigma^2) = \frac{n}{2(\sigma^2)^2} - \frac{2 \cdot \sum_{i=1}^n x_i^2}{2(\sigma^2)^3}$$

$$= \frac{n}{2(\sigma^2)^2} - \frac{\sum_{i=1}^n x_i^2}{(\sigma^2)^3}$$

To compute the Observed Fisher Information we use equation:

$$I(x) = -\ddot{\ell}_x(\sigma^2)$$

$$= \frac{\sum_{i=1}^n x_i^2}{(\sigma^2)^3} - \frac{n}{2(\sigma^2)^2}$$

and substitute  $\sigma^2$  with  $\hat{\sigma}^2_{MLE}$

Thus...



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Thus ...

$$I(\underline{x}) = \frac{\sum_{i=1}^n x_i^2}{\left(\frac{\sum_{i=1}^n x_i^2}{n}\right)^3} - \frac{n}{2 \left(\frac{\sum_{i=1}^n x_i^2}{n}\right)^2}$$

$$= \frac{n^3}{\left(\sum_{i=1}^n x_i^2\right)^2} - \frac{n^3}{2 \left(\sum_{i=1}^n x_i^2\right)^2}$$

$$= \frac{n^3}{2 \left(\sum_{i=1}^n x_i^2\right)^2}$$

c)  $\hat{\sigma}_i^2 \sim N\left(\frac{n-1}{n} \cdot \sigma^2, \frac{1}{I(\underline{x}_i)}\right)$



c)  $\hat{\sigma}_1^2 \sim N\left(\underbrace{\frac{n-1}{n} \cdot \sigma^2}, \frac{1}{I(x_1)}\right)$

recall  $\hat{\sigma}_{MLE}^2$  is biased.

use sample 1 data and this equation.

$$\hat{\sigma}_2^2 \sim N\left(\frac{n-1}{n} \cdot \sigma^2, \frac{1}{I(x_2)}\right)$$

use sample 2 data to compute.