

M378K: November 19th, 2025.

More About Maximum Likelihood Estimation.

Def'n. An estimator $\hat{\Theta} = \hat{\Theta}(y_1, y_2, \dots, y_n)$ is called the **maximum likelihood estimator (MLE)** if it satisfies that for any $\hat{\Theta}' = \hat{\Theta}'(y_1, \dots, y_n)$ we have

$$L(\hat{\Theta}; y_1, \dots, y_n) \geq L(\hat{\Theta}'; y_1, \dots, y_n) \\ \text{for all } y_1, \dots, y_n$$

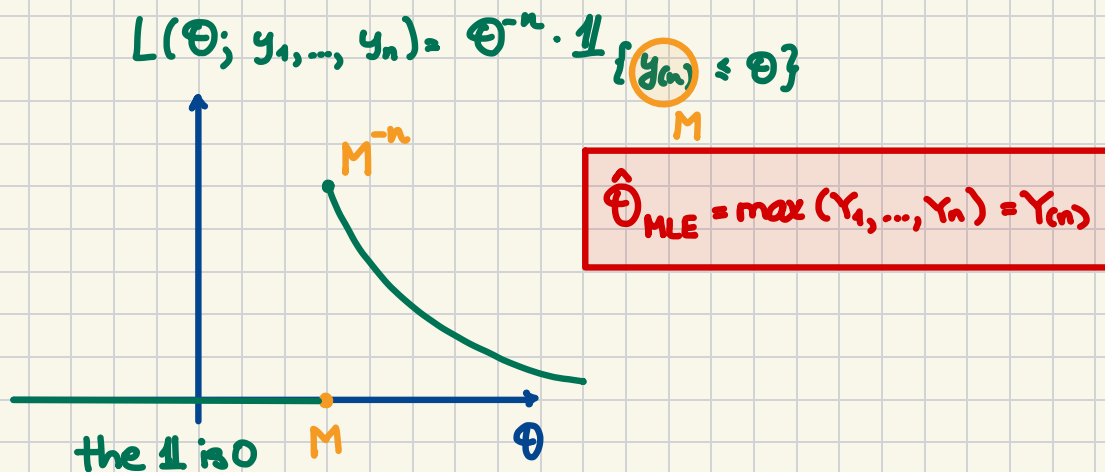
Example. $Y_1, Y_2, \dots, Y_n \sim U(0, \Theta)$ $\Theta > 0$ unknown

The pdf of $U(0, \Theta)$:

$$f^\Theta(y) = \frac{1}{\Theta} \mathbb{1}_{[0, \Theta]}(y) = \frac{1}{\Theta} \mathbb{1}_{\{0 \leq y \leq \Theta\}}$$

The likelihood:

$$\begin{aligned} L(\Theta; y_1, \dots, y_n) &= f^\Theta(y_1) \cdot f^\Theta(y_2) \cdots f^\Theta(y_n) \\ &= \left(\frac{1}{\Theta} \cdot \mathbb{1}_{\{0 \leq y_1 \leq \Theta\}} \right) \cdots \left(\frac{1}{\Theta} \cdot \mathbb{1}_{\{0 \leq y_n \leq \Theta\}} \right) \\ &= \left(\frac{1}{\Theta} \right)^n \mathbb{1}_{\{0 \leq y_1, \dots, y_n \leq \Theta\}} \\ &= \left(\frac{1}{\Theta} \right)^n \mathbb{1}_{\{0 \leq \min(y_1, \dots, y_n)\}} \cdot \mathbb{1}_{\{\max(y_1, \dots, y_n) \leq \Theta\}} \\ &\quad \text{"assume } y_1, \dots, y_n \geq 0\text{"} \\ &= \Theta^{-n} \cdot \mathbb{1}_{\{\max(y_1, \dots, y_n) \leq \Theta\}} \end{aligned}$$



Q: Is this the same as the moment matching estimator?

→: Moment matching:

theoretical mean = sample mean

$$\frac{\Theta}{2} =$$

$$\bar{Y}$$

$$\Rightarrow \hat{\Theta}_{MM} = 2\bar{Y}$$

Example. Let Y_1, \dots, Y_n be a random sample from a dist'n w/ density

$$f^\Theta(y) = \Theta \cdot y^{\Theta-1} \cdot \mathbb{1}_{[0,1]}(y)$$

for some unknown positive parameter Θ .
Find the MLE for Θ .

→:

$$L(\Theta; y_1, \dots, y_n) = \prod_{i=1}^n f^\Theta(y_i)$$

$$= \prod_{i=1}^n (\Theta y_i^{\Theta-1}) = \Theta^n \left(\prod_{i=1}^n y_i \right)^{\Theta-1}$$

$$l(\theta; y_1, \dots, y_n) = n \cdot \ln(\theta) + (\theta - 1) \cdot \ln\left(\prod_{i=1}^n y_i\right)$$

$$l'(\theta; y_1, \dots, y_n) = n \cdot \frac{1}{\theta} + \sum_{i=1}^n \ln(y_i) = 0$$

$$\frac{n}{\theta} = - \sum_{i=1}^n \ln(y_i)$$

$$\hat{\theta}_{MLE} = - \frac{n}{\sum_{i=1}^n \ln(y_i)}$$

□

Example. CAS Exam 3, Spring 2007.

Consider a random sample

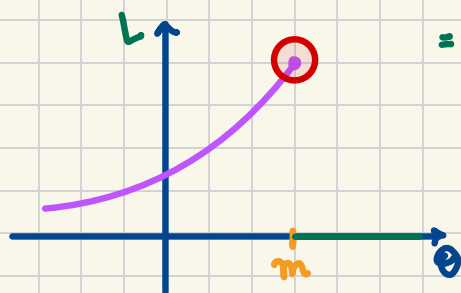
Y_1, \dots, Y_n from a dist'n w/ the pdf

$$f^\theta(y) = e^{-y+\theta} \cdot \mathbb{1}_{(\theta, \infty)}(y).$$

Find the MLE for θ .

$$\rightarrow: L(\theta; y_1, \dots, y_n) = \prod_{i=1}^n \left(e^{-y_i+\theta} \cdot \mathbb{1}_{\{\theta < y_i\}} \right)$$

$$= \left(\prod_{i=1}^n e^{-y_i+\theta} \right) \cdot \mathbb{1}_{\{\theta < \min(y_1, \dots, y_n)\}}$$



$$\hat{\theta}_{MLE} = Y_{(1)}$$

$y_{(1)}$
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