

Elementary Statistical Inference

Binomial distribution ;

Let $X = \begin{cases} 1 & \text{succes} \\ 0 & \text{otherwise} \end{cases}$ $X \sim \text{Bin}(n, \theta), \theta \in [0, 1]$

Bayes estimator (continuous case)

The Bayes estimator w.r.t. the prior density π is the estimator T that minimizes $R(\pi; T)$ over all T .

The Bayes estimate for $g(\theta)$ w.r.t. the posterior density π is given by :

$$\bar{T}(x) = \frac{\int g(\theta) p_{\theta}(x) \pi(\theta) d\theta}{\int p_{\theta}(x) \pi(\theta) d\theta}$$

n known ; $\theta \in [0, 1]$ to estimate ; Beta prior

$$\pi(\theta) = \frac{1}{B(\alpha, \beta)} \theta^{\alpha-1} (1-\theta)^{\beta-1} \mathbb{I}_{[0,1]}(\theta)$$

We find that $\hat{\theta}_{\text{bay}} = T_{\alpha, \beta}(X) = \frac{X + \alpha}{n + \alpha + \beta}$ with $X = \sum_{i=1}^n X_i$

We find a different Bayes estimator for each combination of parameters (α, β) ; $\alpha > 0$; $\beta > 0$.

Note : $\hat{\theta}_{\text{MLE}} = X/n$ is not in the class of Bayes estimators. It is the limit case $(\alpha, \beta) \rightarrow (0, 0)$

Reference : An introduction to mathematical statistics,
Bijma ; Jonker ; van der Vaart , 2016