

O erro quadrático médio (EQM) do **Estimador**  $T(X_1, \dots, X_n)$  com respeito a  $g(\theta)$  é definido por

$$\text{EQM}(T, g(\theta)) = E_{\theta}((T(X_1, \dots, X_n) - g(\theta))^2)$$

Obs.:

Se  $T(X_1, \dots, X_n)$  for não viciado para  $g(\theta)$ , então

$$\text{EQM}(T, g(\theta)) = \text{Var}_{\theta}(T(X_1, \dots, X_n)) \forall \theta \in \Theta$$

## Propriedades do EQM

Seja  $T(X_1, \dots, X_n)$  um estimador para  $g(\theta)$ , seja  $\mu_t = E_{\theta}(T(X_1, \dots, X_n))$

$$\begin{aligned} \text{EQM}(T, g(\theta)) &= E_{\theta}[(T(X_1, \dots, X_n) - \mu_t + \mu_t - g(\theta))^2] \\ &= E_{\theta}[((T(X_1, \dots, X_n) - \mu_t) + (\mu_t - g(\theta)))^2] \\ &= E_{\theta}[(T(X_1, \dots, X_n) - \mu_t)^2 + 2(T(X_1, \dots, X_n) - \mu_t)(\mu_t - g(\theta)) + (\mu_t - g(\theta))^2] \\ &= \underbrace{E_{\theta}[(T(X_1, \dots, X_n) - \mu_t)^2]}_{\text{Var}_{\theta}(T(X_1, \dots, X_n))} + 2(\mu_t - g(\theta)) \underbrace{E_{\theta}(T(X_1, \dots, X_n) - \mu_t)}_0 + (\mu_t - g(\theta))^2 \\ &= \text{Var}_{\theta}(T(X_1, \dots, X_n)) + (\mu_t - g(\theta))^2 \end{aligned}$$

Portanto,

$$\text{EQM}(T, g(\theta)) = \text{Var}_{\theta}(T(X_1, \dots, X_n)) + (\mu_t - g(\theta))^2$$

## Viés

Denotamos de viés de  $T(X_1, \dots, X_n)$  com respeito a  $g(\theta)$  por

$$\text{Viés}(T, g(\theta)) = E_{\theta}(T(X_1, \dots, X_n)) - g(\theta), \forall \theta \in \Theta$$

Dessa forma, temos que

$$\text{EQM}(T, g(\theta)) = \text{Var}_{\theta}(T(X_1, \dots, X_n)) + [\text{Viés}(T, g(\theta))]^2$$

## Exemplo

Seja  $(X_1, \dots, X_n)$  uma **amostra aleatória**, ou seja, independentes e identicamente distribuídas (iid), de  $X \sim \text{Ber}(\theta)$  em que  $\theta \in \Theta = (0, 1)$ . Calcule o viés e o EQM de  $\bar{X}_n$  com respeito a  $g(\theta) = P_{\theta}(X = 1)$

O estimador é, então,  $T(X_1, \dots, X_n) = \bar{X}_n = \frac{X_1 + \dots + X_n}{n}$  para  $g(\theta) = P_{\theta}(X = 1) = \theta$  (pelo **modelo de Bernoulli**).

$$\begin{aligned}
E_{\theta}(\bar{X}_n) &= E_{\theta} \left( \frac{1}{n} \sum_{i=1}^n X_i \right) = \frac{1}{n} \sum_{i=1}^n E_{\theta}(X_i) \stackrel{id. dist.}{\Rightarrow} \\
E_{\theta}(\bar{X}_n) &= \frac{1}{n} \sum_{i=1}^n E_{\theta}(X), \forall \theta \in \Theta \\
&= \frac{n}{n} \theta = \theta, \forall \theta \in \Theta
\end{aligned}$$

Portanto,  $\bar{X}_{\theta}$  é não enviesado para  $g(\theta) = \theta$ .

$$\Rightarrow \text{Viés}(\bar{X}_n, g(\theta)) = 0, \forall \theta \in \Theta$$

Para o EQM,

$$\begin{aligned}
\text{EQM}(\bar{X}_n, g(\theta)) &= \text{Var}_{\theta}(\bar{X}_n) - 0^2 = \text{Var}_{\theta} \left( \frac{1}{n} \sum_{i=1}^n X_i \right) = \frac{1}{n^2} \text{Var}_{\theta} \left( \sum_{i=1}^n X_i \right) \\
&\stackrel{\text{ind}}{=} \frac{1}{n^2} \sum_{i=1}^n \text{Var}_{\theta}(X_i) \stackrel{\text{ind. dist.}}{=} \frac{1}{n^2} \sum_{i=1}^n \text{Var}_{\theta}(X), \forall \theta \in \Theta \\
&= \frac{n\theta(1-\theta)}{n^2} = \frac{\theta(1-\theta)}{n}, \forall \theta \in \Theta
\end{aligned}$$