

## TA section 4

JERRY C.

108354501@nccu.edu.tw

April 23, 2024

## Review: Point Estimation

# 估計量的優劣評估

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–  $q(\theta)$  之均方差 (MSE):

$$MSE_{\theta}(T) = R(\theta, T) = \mathbb{E}[(T(\mathbf{X}) - q(\theta))^2] = \text{Var}[T(\mathbf{X})] + Bias_{\theta}^2(T(\mathbf{X})).$$

–  $\text{Var}[T(\mathbf{X})] = \mathbb{E}[(T(\mathbf{X}) - \mathbb{E} T(\mathbf{X}))^2]$ , 且  $Bias_{\theta}(T(\mathbf{X})) = \mathbb{E}[T(\mathbf{X})] - q(\theta)$ .

- **最佳不偏估計量 (best unbiased estimator, BUE):** CRLB (Cramer-Rao lower bound):  $(q'(\theta))^2/I(\theta)$ .
- **一致最小變異不偏估計量 (uniformly minimum variance unbiased estimator, UMVUE):**
  - [Rao-Blackwell Theorem]: (充份 + 不偏  $\Rightarrow$  效率)
  - [Lehmann-Scheffé (-Rao-Blackwell) Theorem]: (完備 + 充份 + 不偏  $\Rightarrow$  效率)
- **BUE  $\Rightarrow$  UMVUE.** ,i.e.,  $\text{Var}[\text{BUE}] \leq \text{Var}[\text{UMVUE}]$ .

# 大樣本性質

## ■ 一致性:

- (weak consistency)  $T_n \xrightarrow{P} q(\theta)$  as  $n \rightarrow \infty$  if

$$\lim_{n \rightarrow \infty} \mathbb{P}(|T_n - q(\theta)| < \epsilon) = 1, \forall \epsilon > 0.$$

- (mean-squares consistency/ $L_2$ -norm consistency)  $T_n \xrightarrow{\text{m.s.}} q(\theta)$  as  $n \rightarrow \infty$  if

$$\mathbb{E}[(T_n - q(\theta))^2] \rightarrow 0, \text{ as } n \rightarrow \infty.$$

或  $\lim_{n \rightarrow \infty} \text{Var}[T_n] = 0$ , 且  $\lim_{n \rightarrow \infty} \text{Bias}_\theta(T_n) = 0$ .

- 漸近不偏性:  $\lim_{n \rightarrow \infty} \mathbb{E}[T_n] = q(\theta)$ .
- 漸近效率性:  $\text{are}(T_{2n}, T_{1n}) = \lim_{n \rightarrow \infty} [\text{Var}[T_{1n}] / \text{Var}[T_{2n}]] \leq 1$  for  $T_{1n}, T_{2n}$  two (asymptotic) unbiased estimators, 則  $T_{1n}$  具漸近有效。
- 漸近常態性: 給定一估計量, 其漸近變異數  $\text{Avar}[T_n] = \sigma_n^2(\theta)$  與漸近期望值  $\text{Asy. } \mathbb{E}[T_n] = \mu_n(\theta)$ , 則  $(T_n - \mu_n(\theta)) / \sigma_n(\theta) \xrightarrow{d} \mathcal{N}(0, 1)$ .
- 最佳漸近常態性: 若  $n\sigma_n^2(\theta) \rightarrow v^2(\theta)$  且  $n^{1/2}(\mu_n(\theta) - q(\theta)) \rightarrow 0$ , 其中  $v^2(\theta) > 0$  為某一  $\theta$  之函數, 則稱  $T_n$  具有最佳漸近常態性。即, 此時漸近變異數為  $v^2(\theta)/n$ 。



# Admissible Estimator

- (inadmissible, 不可行的/不可採用的): 對二估計量  $S, T$ , 若

$$R(\theta, T) \leq R(\theta, S), \forall \theta \in \Omega$$

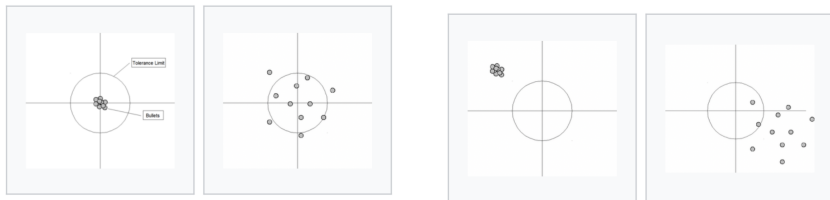
且

$$R(\theta, T) < R(\theta, S), \text{ for some } \theta,$$

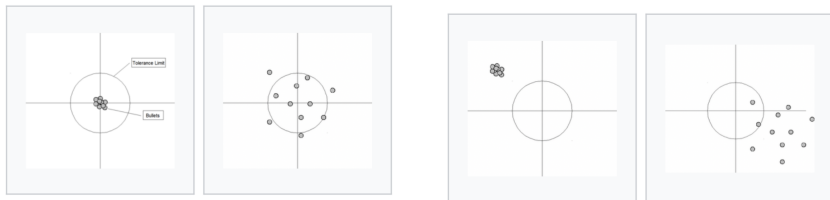
則  $T$  較  $S$  為佳, 且  $S$  為不可行的/不可採用的。

- 若一個估計量  $T$ , 找不到較其為佳的估計量 (找不到均方差更小), 則稱  $T$  可行的/可採用的。
- 若  $S$  為不可行的, 不代表  $T$  是可行的。

# bias-variance tradeoff

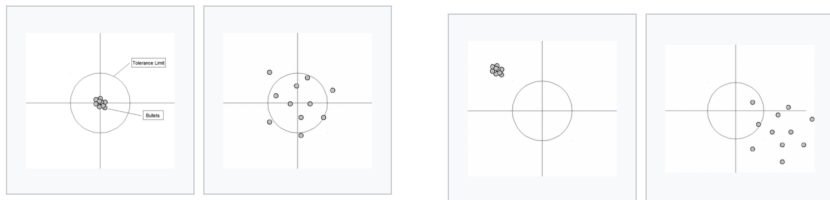


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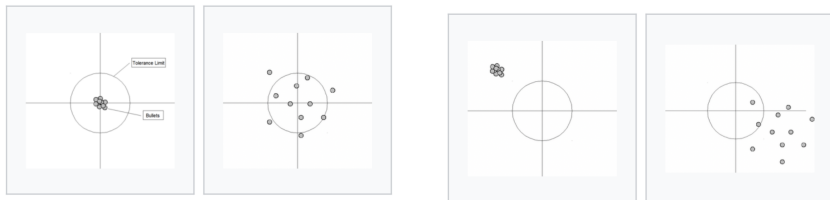
- (1) low bias, low variance;

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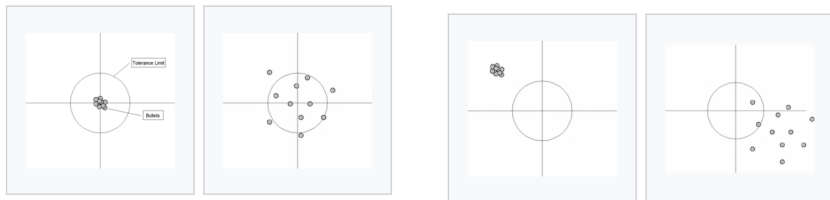
- (1) low bias, low variance;
- (2) low/moderate bias, high variance;

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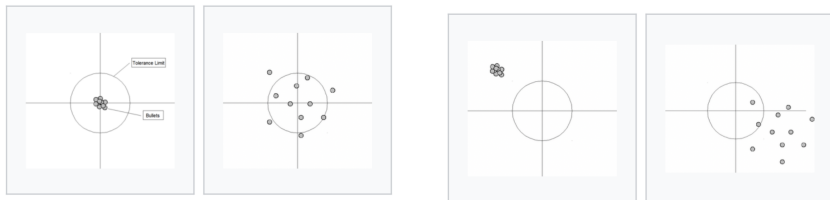
- (1) **low** bias, **low** variance;
- (2) **low/moderate** bias, **high** variance;
- (3) **high** bias, **low** variance;

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- (1) low bias, low variance;
- (2) low/moderate bias, high variance;
- (3) high bias, low variance;
- (4) high bias, high variance.

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# Example

- $X_1, \dots, X_n \sim \mathcal{N}(\mu, \sigma^2)$ . Let  $q(\sigma^2) := \sigma^2$ , consider two estimators of  $q(\sigma^2)$ ,  
 $T_1 = \sum_{i=1}^n (X_i - \bar{X})^2/n$ ,  $T_2 = \sum_{i=1}^n (X_i - \bar{X})^2/(n-1) = nT_1/(n-1)$ .
- $\mathbb{E}[T_1] = (n-1)\sigma^2/n \neq \sigma^2$  (biased for  $\sigma^2$ ),  $\text{Var}[T_1] = 2(n-1)\sigma^4/n^2$ . So,

$$R(\theta, T_1) = \text{Var}[T_1] + \text{Bias}^2(\theta, T_1) = \frac{2(n-1)\sigma^4}{n^2} + \frac{\sigma^4}{n^2} = \frac{(2n-1)\sigma^4}{n^2}.$$

- $\mathbb{E}[T_2] = \sigma^2$  (unbiased for  $\sigma^2$ ),  $\text{Var}[T_2] = 2\sigma^4/(n-1) > \text{Var}[T_1]$ . So,

$$R(\theta, T_2) = 2\sigma^4/(n-1) > R(\theta, T_1).$$

- $T_2$  不偏, 但 MSE 卻較大 (因變異數大的幅度超過  $T_1$  偏誤的幅度)
- $\lim_{n \rightarrow \infty} \mathbb{E}[T_1] = \lim_{n \rightarrow \infty} \mathbb{E}[T_2] = \sigma^2$  (漸近不偏),  
 $\lim_{n \rightarrow \infty} \text{Var}[T_1] = \lim_{n \rightarrow \infty} \text{Var}[T_2] = 0$  (均方差一致性);

$$\text{are}(T_2, T_1) = \lim_{n \rightarrow \infty} \text{Var}[T_1]/\text{Var}[T_2] = \lim_{n \rightarrow \infty} \frac{2(n-1)\sigma^4/n^2}{2\sigma^4/(n-1)} = 1,$$

且  $T_2$  是 UMVUE, 故  $T_2$  具漸近有效性。

- (若不考慮不偏性)

$$\text{Var}[T_1] < \text{CRLB}(\sigma^2) = 2\sigma^4/n < \text{Var}[T_2].$$