

Lecture 07. Statistical Inference (I):

Point- and Interval Estimation

統計推論 (I)：點估計與區間估計

- 信賴區間(confidence interval, CI)
- 點估計 point estimation ; 點估計值 point estimate
- 區間估計 interval estimation
- Sampling distribution of a statistic \Leftrightarrow parameter
- Properties of an estimator
 - Unbiasedness
 - Consistency
- Two-sided (two-tailed) CI (• One-sided (one-tailed) CI)
- Confidence interval for the population mean with NORMALITY assumption;
 - 1. variance known
 - 2. variance unknown: The Student's t-statistic
 - Graphical illustrations for the meaning of confidence interval
- Confidence interval for the population variance with NORMALITY assumption;
- Confidence interval for the population mean without NORMALITY assumption
- EXAMPLES

常態分布平均值(期望值) μ 之點估計與信賴區間

If $X_1, X_2, \dots, X_n \sim N(\mu, \sigma^2)$,

Q: How to estimate μ ?

A: $\hat{\mu} = \bar{X}$

We call \bar{X} a **point estimator of μ** (點估計式)。

The **realization of \bar{X}** , $\bar{x} = (x_1 + x_2 + \dots + x_n)/n$, is the **point estimate of μ** (點估計值)。

PROPERTIES

1. Unbiasedness (不偏性)

$$E(\bar{X}) = \mu$$

$$\begin{aligned} E(\bar{X}) &= E[(X_1 + X_2 + \dots + X_n)/n] \\ &= (1/n)[E(X_1) + E(X_2) + \dots + E(X_n)] \\ &= (1/n)[\mu + \mu + \dots + \mu] = \mu \end{aligned}$$

2. Consistency (趨近性；一致性) ★

$\bar{X} \rightarrow \mu$, in probability, or

$\Pr(|\bar{X} - \mu| < \varepsilon) > 1 - \varepsilon$, for arbitrary (usually small) $\varepsilon > 0$

Weak Law of Large Number (弱大數法則)

Sampling distribution of the sample mean

(under 'NORMALITY' assumption) : (I)

- If $X_1, X_2, \dots, X_n \sim N(\mu, \sigma^2)$, σ is a **known** value

$$\implies \boxed{\bar{X} \sim N(\mu, \sigma^2/n)} \quad (1)$$

- That is, the sampling distribution of \bar{X} is still Gaussian (normal), for all **n**.

- A suitable standardization [Q1] for (1) gives

$$\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim N(0, 1). \quad (2)$$

If σ is assumed to be known (but usually it is not), (2)

implies:

$$\Pr(-1.96 < \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} < 1.96) = 0.95,$$

$$\Pr(-1.645 < \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} < 1.645) = 0.90, \dots$$

Q1: If $Y \sim N(\mu, \sigma^2)$, $\Rightarrow Y - \mu \sim N(0, \sigma^2)$

$$\Rightarrow Y - \mu / \sigma \sim N(0, 1) \text{ [why??]}$$

$$\Leftarrow \text{Var}[Y - \mu / \sigma] = \text{Var}[Y - \mu] / \sigma^2$$

$$= \text{Var}[Y] / \sigma^2 = 1.$$

$$\Pr(-Z_{\alpha/2} < \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} < Z_{\alpha/2}) = 1 - \alpha, \text{ or } = 100(1 - \alpha)\%$$

$\alpha=0.05$, $Z_{0.025}=1.96$; $\alpha=0.10$, $Z_{0.05}=1.645$;...

We have a two-sided confidence interval

$$\Pr(\bar{X} - Z_{\alpha/2} \frac{\sigma}{\sqrt{n}} < \mu < \bar{X} + Z_{\alpha/2} \frac{\sigma}{\sqrt{n}}) = 1 - \alpha \quad (3)$$

■ Conclusions

$$\mu \text{ 之 } 100(1-\alpha)\% \text{ CI} = (\bar{X} - Z_{\alpha/2} \frac{\sigma}{\sqrt{n}} , \bar{X} + Z_{\alpha/2} \frac{\sigma}{\sqrt{n}}) \quad (3')$$

$$\mu \text{ 之 } 95\% \text{ CI} = (\bar{X} - 1.96 \frac{\sigma}{\sqrt{n}} , \bar{X} + 1.96 \frac{\sigma}{\sqrt{n}})$$

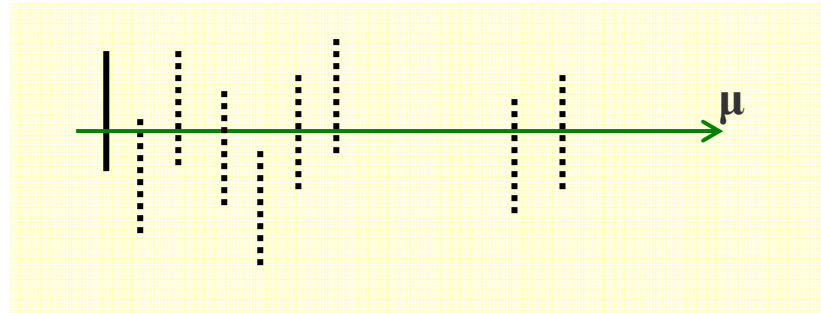
$$\mu \text{ 之 } 90\% \text{ CI} = (\bar{X} - 1.645 \frac{\sigma}{\sqrt{n}} , \bar{X} + 1.645 \frac{\sigma}{\sqrt{n}})$$

...

confidence limits

■ 信賴區間的解釋 (very important !!)

- Graphical illustrations for the meaning of confidence interval



- If thus constructed 95% CI can have infinite numbers of realizations, then, in average, they will cover the true parameter close to 95 times in 100 replicates. (In a long run !!) (Frequentist's point of view!!)

Q. Why is it necessary to understand the property of confidence interval in the above statement when in practice you will only have a single realization?

■ Q and A

- 在同樣的信心水準(confidence)下，不同的統計方法可能得到不同的信賴區間。不同的信賴區間彼此間是否可以互相比較？
- 信賴區間的長度是否越短越好？
- 信賴區間的長度決定於哪些因素？為什麼？

Sampling distribution of the sample mean

(under 'NORMALITY' assumption) : (II)

Variance (σ^2) **unknown** and the Student's t-statistic

When σ (or σ^2) is unknown, it is not reasonable to report the 95%CI (say) of μ in terms of σ . Instead, the σ appeared in (3) should be replaced by an estimate of σ . Conventionally, we use $s = \hat{\sigma}$, where $s^2 = \Sigma(x_i - \bar{x})^2 / (n-1)$.

Question:

$$\frac{\bar{x} - \mu}{\sigma / \sqrt{n}} \sim \mathbf{N}(0, 1) \implies \frac{\bar{x} - \mu}{\hat{\sigma} / \sqrt{n}} \sim \mathbf{N}(0, 1)???$$

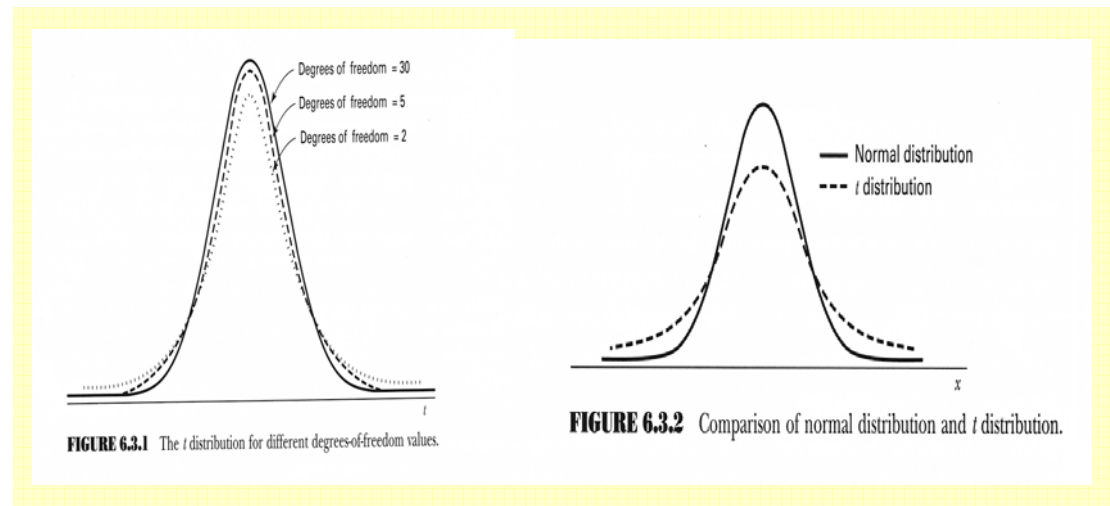
Answer : ★

$$\begin{aligned} \frac{\bar{x} - \mu}{\hat{\sigma} / \sqrt{n}} &= \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} \cdot \frac{1}{\frac{\hat{\sigma}}{\sigma}} \\ &= \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} \cdot \frac{1}{\sqrt{\frac{\sum (x_i - \bar{x})^2 / \sigma^2}{n-1}}} \\ &= \mathbf{Z} \cdot \frac{1}{\sqrt{\frac{\chi_{n-1}^2}{n-1}}} \equiv \mathbf{t}_{n-1}, \end{aligned}$$

$$\text{where } \mathbf{t}_{\nu}^2 \equiv \frac{\chi_1^2 / 1}{\chi_{\nu}^2 / \nu}$$

□ The results were derived and published by W. S. Gosset (1908, Biometrika) with the pseudonym of ‘Student’.

□ **Characteristics**



□ **Confidence intervals for μ (say, $n=21$):**

μ 之 $100(1-\alpha)\%$ CI=

$$\left(\bar{X} - t_{n-1, \alpha/2} \frac{s}{\sqrt{n}} , \bar{X} + t_{n-1, \alpha/2} \frac{s}{\sqrt{n}} \right) \quad (4)$$

95% CI

$$\left(\bar{X} - 2.086 \frac{s}{\sqrt{n}} , \bar{X} + 2.086 \frac{s}{\sqrt{n}} \right)$$

90% CI

$$\left(\bar{X} - 1.725 \frac{s}{\sqrt{n}} , \bar{X} + 1.725 \frac{s}{\sqrt{n}} \right)$$

□ **Example: (See the textbook.)**

Confidence interval for the population mean without NORMALITY assumption: an application of CLT

$X_1, X_2, \dots, X_n \sim F(\bullet)$, an unknown distribution with $E(X)=\mu$ and $\text{Var}(X)=\sigma^2$. Here we derive the **approximate 100(1- α) % CI** of μ :

since $\bar{X} \sim N(\mu, \sigma^2/n)$, or $\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim N(0, 1)$

The approximate CI is

$(\bar{X} - Z_{\alpha/2} \frac{\sigma}{\sqrt{n}}, \bar{X} + Z_{\alpha/2} \frac{\sigma}{\sqrt{n}})$; when n gets larger, the approximation

gets better. However, since σ is usually unknown, it is replaced by s , the sample standard deviation; in this case, the **approximate CI** is

$$\left(\bar{X} - Z_{\alpha/2} \frac{s}{\sqrt{n}}, \bar{X} + Z_{\alpha/2} \frac{s}{\sqrt{n}} \right)$$

■ **Example:** Tossing a coin 100 times, and you get 45 ‘heads’. Let p be the probability of getting a ‘head’. Please calculate the 95% CI of p .

Ans: $X_1 \dots X_{100} \sim \text{Ber}(p)$, each $X_i=0$ (tail) or 1 (head). By CLT,

$$\frac{\bar{x} - p}{\sqrt{\text{Var}(\bar{x})}} \sim N(0, 1)$$

. If we use

$$\widehat{\text{Var}}(\bar{x}) = \frac{\hat{p}(1 - \hat{p})}{n}$$

, then

$$\Pr \left\{ \hat{p} - z_{\alpha/2} \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}} \leq p \leq \hat{p} + z_{\alpha/2} \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}} \right\} = 1 - \alpha$$

So, setting $\alpha=0.05$ leads to the 95% CI of p :

$(0.45 - 1.96 * \text{sqrt}(0.45 * 0.55 / 100), 0.45 + 1.96 * \text{sqrt}(0.45 * 0.55 / 100))$

$= (0.352, 0.548)$