

## Chapter 5. Confidence Interval

Let  $Z$  be the r.v. with standard normal distribution  $N(0, 1)$

We can find  $z_\alpha$  and  $z_{\frac{\alpha}{2}}$  that satisfy

$$\alpha = P(Z \leq -z_\alpha) = P(Z \geq z_\alpha) \text{ and } 1 - \alpha = P(-z_{\frac{\alpha}{2}} \leq Z \leq z_{\frac{\alpha}{2}}).$$

A table of  $z_{\frac{\alpha}{2}}$  is the following :

$1 - \alpha$	$z_{\frac{\alpha}{2}}$
0.8	1.28 ( $z_{0.1}$ )
0.9	1.645 ( $z_{0.05}$ )
0.95	1.96 ( $z_{0.025}$ )
0.99	2.58 ( $z_{0.005}$ )
0.9973	3 ( $z_{0.00135}$ )

**Def.** Suppose that we have a random sample from  $f(x, \theta)$ . For  $0 < \alpha < 1$ , if there exists two statistics  $T_1 = t_1(X_1, \dots, X_n)$  and  $T_2 = t_2(X_1, \dots, X_n)$  satisfying

$$1 - \alpha = P(T_1 \leq \theta \leq T_2)$$

We call the random interval  $(T_1, T_2)$  a  $100(1 - \alpha)\%$  confidence interval of parameter  $\theta$ . If  $X_1 = x_1, \dots, X_n = x_n$  is observed, we also call  $(t_1(X_1, \dots, X_n), t_2(X_1, \dots, X_n))$  a  $100(1 - \alpha)\%$  confidence interval (C.I.) for  $\theta$

Constructing C.I. by pivotal quantity:

**Def.** A function of random sample and parameter,  $Q = q(X_1, \dots, X_n, \theta)$ , is called a pivotal quantity if its distribution is independent of  $\theta$

With a pivotal quantity  $q(X_1, \dots, X_n, \theta)$ , there exists  $a, b$  such that

$$1 - \alpha = P(a \leq q(X_1, \dots, X_n, \theta) \leq b), \forall \theta \in \Theta.$$

The interest of pivotal quantity is that there exists statistics  $T_1 = t_1(X_1, \dots, X_n)$  and  $T_2 = t_2(X_1, \dots, X_n)$  with the following 1-1 transformation

$$a \leq q(X_1, \dots, X_n, \theta) \leq b \text{ iff } T_1 \leq \theta \leq T_2$$

Then we have  $1 - \alpha = P(T_1 \leq \theta \leq T_2)$  and  $(T_1, T_2)$  is a  $100(1 - \alpha)\%$  C.I. for  $\theta$

Confidence Interval for Normal mean:

Let  $X_1, \dots, X_n$  be a random sample from  $N(\mu, \sigma^2)$ . We consider the C.I. of

parameter  $\mu$ .

(I)  $\sigma = \sigma_0$  is known

$$\bar{X} \sim N\left(\mu, \frac{\sigma_0^2}{n}\right) \Rightarrow \frac{\bar{X} - \mu}{\sigma_0/\sqrt{n}} \sim N(0, 1)$$

$$\begin{aligned} 1 - \alpha &= P(-z_{\frac{\alpha}{2}} \leq Z \leq z_{\frac{\alpha}{2}}), Z \sim N(0, 1) \\ &= P(-z_{\frac{\alpha}{2}} \leq \frac{\bar{X} - \mu}{\sigma_0/\sqrt{n}} \leq z_{\frac{\alpha}{2}}) \\ &= P(-z_{\frac{\alpha}{2}} \frac{\sigma_0}{\sqrt{n}} \leq \bar{X} - \mu \leq z_{\frac{\alpha}{2}} \frac{\sigma_0}{\sqrt{n}}) \\ &= P(\bar{X} - z_{\frac{\alpha}{2}} \frac{\sigma_0}{\sqrt{n}} \leq \mu \leq \bar{X} + z_{\frac{\alpha}{2}} \frac{\sigma_0}{\sqrt{n}}) \end{aligned}$$

$\Rightarrow (\bar{X} - z_{\frac{\alpha}{2}} \frac{\sigma_0}{\sqrt{n}}, \bar{X} + z_{\frac{\alpha}{2}} \frac{\sigma_0}{\sqrt{n}})$  is a  $100(1 - \alpha)\%$  C.I. for  $\mu$ .

ex:  $n = 40, \sigma_0 = \sqrt{10}, \bar{x} = 7.164$  ( $X_1, \dots, X_{40} \stackrel{iid}{\sim} N(\mu, 10)$ .)

Want a 80% C.I. for  $\mu$ .

sol: A 80% C.I. for  $\mu$  is

$$\begin{aligned} \left(\bar{X} - z_{\frac{\alpha}{2}} \frac{\sigma_0}{\sqrt{n}}, \bar{X} + z_{\frac{\alpha}{2}} \frac{\sigma_0}{\sqrt{n}}\right) &= \left(7.164 - 1.28 \frac{\sqrt{10}}{\sqrt{40}}, 7.164 + 1.28 \frac{\sqrt{10}}{\sqrt{40}}\right) \\ &= (6.523, 7.805) \end{aligned}$$

$$P\left(\bar{X} - z_{\frac{\alpha}{2}} \frac{\sigma_0}{\sqrt{n}} \leq \mu \leq \bar{X} + z_{\frac{\alpha}{2}} \frac{\sigma_0}{\sqrt{n}}\right) = 1 - \alpha = 0.8$$

$$P(6.523 \leq \mu \leq 7.805) = 1 \text{ or } 0$$

(II)  $\sigma$  is unknown.

**Def.** If  $Z \sim N(0, 1)$  and  $\chi^2(r)$  are independent, we call the distribution of the r.v.

$$T = \frac{Z}{\sqrt{\frac{\chi^2(r)}{r}}}$$

a t-distribution with  $r$  degrees of freedom.

The p.d.f of t-distribution is

$$f_T(t) = \frac{\Gamma(\frac{r+1}{2})}{\Gamma(\frac{r}{2})} \frac{1}{\sqrt{r\pi} \left(1 + \frac{t^2}{r}\right)^{\frac{r+1}{2}}}, -\infty < t < \infty$$

$\because f_T(-t) = f_T(t)$

$\therefore$  t-distribution is symmetric at 0.

Now  $X_1, \dots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$ . We have

$$\left\{ \begin{array}{l} \bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right) \\ \frac{(n-1)s^2}{\sigma^2} \sim \chi^2(n-1) \end{array} \right. \text{ indep.} \Rightarrow \left\{ \begin{array}{l} \frac{\bar{X}-\mu}{\sigma/\sqrt{n}} \sim N(0, 1) \\ \frac{(n-1)s^2}{\sigma^2} \sim \chi^2(n-1) \end{array} \right. \text{ indep.}$$

$$T = \frac{\frac{\bar{X}-\mu}{\sigma/\sqrt{n}}}{\sqrt{\frac{(n-1)s^2}{\sigma^2(n-1)}}} = \frac{\bar{X} - \mu}{s/\sqrt{n}} \sim t(n-1)$$

Let  $t_{\frac{\alpha}{2}}$  satisfies

$$\begin{aligned} 1 - \alpha &= P(-t_{\frac{\alpha}{2}} \leq \frac{\bar{X} - \mu}{s/\sqrt{n}} \leq t_{\frac{\alpha}{2}}) \\ &= P(-t_{\frac{\alpha}{2}} \frac{s}{\sqrt{n}} \leq \bar{X} - \mu \leq t_{\frac{\alpha}{2}} \frac{s}{\sqrt{n}}) \\ &= P(\bar{X} - t_{\frac{\alpha}{2}} \frac{s}{\sqrt{n}} \leq \mu \leq \bar{X} + t_{\frac{\alpha}{2}} \frac{s}{\sqrt{n}}) \end{aligned}$$

$\Rightarrow (\bar{X} - t_{\frac{\alpha}{2}} \frac{s}{\sqrt{n}}, \bar{X} + t_{\frac{\alpha}{2}} \frac{s}{\sqrt{n}})$  is a  $100(1 - \alpha)\%$  C.I. for  $\mu$ .

ex: Suppose that we have  $n = 10, \bar{x} = 3.22$  and  $s = 1.17$ . We also have  $t_{0.025} = 2.262$ . Want a 95% C.I. for  $\mu$ .

sol: A 95% C.I. for  $\mu$  is

$$\left( 3.22 - 2.262 \frac{1.17}{\sqrt{10}}, 3.22 + 2.262 \frac{1.17}{\sqrt{10}} \right) = (2.34, 4.10)$$