

## Using pivots to construct confidence intervals

In Example 41 we used the fact that

$$Q(\bar{X}, \mu) = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim N(0, 1) \quad \text{for all } \mu.$$

We then said  $|Q(\bar{X}, \mu)| \leq z_{\alpha/2}$  with probability  $1 - \alpha$ , and converted this into a statement about  $\mu$ .

**Definition 21** Given a data vector  $\mathbf{X}$ , a random variable  $Q(\mathbf{X}, \theta)$  is a *pivotal quantity* if the distribution of  $Q(\mathbf{X}, \theta)$  is independent of all unknown parameters.

Suppose we want a  $(1 - \alpha)100\%$  confidence interval for  $\theta$ . Try to find a function of the data that also depends on  $\theta$  but whose probability distribution *does not* depend on  $\theta$ .

Ideally, the random variable  $Q(\mathbf{X}, \theta)$  will depend on  $\mathbf{X}$  only through a sufficient statistic.

Let  $\alpha_1$  and  $\alpha_2$  be positive and such that  $\alpha_1 + \alpha_2 = \alpha$ . Let  $q_1$  and  $q_2$  be the  $\alpha_1$  and  $1 - \alpha_2$  quantiles of the distribution of  $Q(\mathbf{X}, \theta)$ . Then

$$P(q_1 < Q(\mathbf{X}, \theta) \leq q_2) = 1 - \alpha.$$

Now, try to find  $C(\mathbf{X})$  such that

$$\theta \in C(\mathbf{X}) \quad \text{iff} \quad q_1 < Q(\mathbf{X}, \theta) \leq q_2.$$

It then follows that  $C(\mathbf{X})$  is a  $(1 - \alpha)100\%$  confidence set for  $\theta$ .

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**Example 42** Let  $X_1, \dots, X_n$  be a random sample from the  $U(0, \theta)$  distribution. Would like to construct a  $(1 - \alpha)100\%$  confidence interval for  $\theta$ .

What kind of transformation of the data would make the distribution of the transformed data free of  $\theta$ ?

$$\frac{X_1}{\theta}, \dots, \frac{X_n}{\theta} \quad \text{are i.i.d. } U(0, 1).$$

It follows that  $Q(\mathbf{X}, \theta) = \max(X_1/\theta, \dots, X_n/\theta)$  has the same distribution as the maximum of a random sample from the  $U(0, 1)$  distribution. Therefore,  $Q(\mathbf{X}, \theta)$  is a pivotal quantity. Note that

$$Q(\mathbf{X}, \theta) = \frac{X_{(n)}}{\theta},$$

and thus depends on the data through a sufficient statistic.

Choose  $a$  and  $b$  so that

$$P\left(a < \frac{X_{(n)}}{\theta} < b\right) = 1 - \alpha.$$

$$P\left(a < \frac{X_{(n)}}{\theta} < b\right) = \int_a^b nt^{n-1} dt = b^n - a^n.$$

Since

$$P\left(a < \frac{X_{(n)}}{\theta} < b\right) = P\left(\frac{X_{(n)}}{b} < \theta < \frac{X_{(n)}}{a}\right),$$

$[X_{(n)}/b, X_{(n)}/a]$  is a  $(1 - \alpha)100\%$  confidence interval for  $\theta$  so long as  $0 < a, b < 1$  and  $b^n - a^n = 1 - \alpha$ .

It would be desirable to find an interval of the above form that has the shortest possible length. The length is

$$X_{(n)} \left(\frac{1}{a} - \frac{1}{b}\right).$$

We have no control over  $X_{(n)}$ , but we could choose  $a$  and  $b$  to minimize  $a^{-1} - b^{-1}$  subject to the constraint  $b^n - a^n = 1 - \alpha$ .

It is straightforward to show that the solution to the previous problem is

$$b = 1 \quad \text{and} \quad a = \alpha^{1/n}.$$

Therefore, among  $(1 - \alpha)100\%$  confidence intervals of the form  $[X_{(n)}/b, X_{(n)}/a]$ , the shortest one is  $[X_{(n)}, X_{(n)}/\alpha^{1/n}]$ .

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**Example 43** Let  $X_1, \dots, X_n$  be a random sample from a density of the form

$$\frac{1}{\sigma} f\left(\frac{x - \mu}{\sigma}\right),$$

where  $f$  is *known*. The parameter space is

$$\Theta = \{(\mu, \sigma) : -\infty < \mu < \infty, \sigma > 0\}.$$

The family of densities

$$\left\{ \frac{1}{\sigma} f\left(\frac{x - \mu}{\sigma}\right) : (\mu, \sigma) \in \Theta \right\}$$

is called a *location-scale family*.

Define

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

and

$$S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2.$$

The random variable  $(\bar{X} - \mu)/S$  is a pivotal quantity, a fact which leads to a confidence interval for  $\mu$ .

**Proof:** Consider

$$\begin{aligned} X_i &= (X_i - \mu) + \mu \\ &= \sigma(X_i - \mu)/\sigma + \mu \\ &= \sigma Z_i + \mu \end{aligned}$$

We then have  $\bar{X} - \mu = \sigma \bar{Z}$ , where

$$\bar{Z} = n^{-1} \sum_{i=1}^n Z_i.$$

$$\begin{aligned}
S^2 &= \frac{1}{n-1} \sum_{i=1}^n (\sigma Z_i - \sigma \bar{Z})^2 \\
&= \sigma^2 S_Z^2.
\end{aligned}$$

It follows that

$$\frac{(\bar{X} - \mu)}{S} = \frac{\bar{Z}}{S_Z}.$$

It is easy to verify that  $Z_i$  has density  $f$ , which is free of any unknown parameters, and the result follows.

The  $N(\mu, \sigma^2)$  family of distributions is an example of a location-scale family, with

$$f(y) = \frac{1}{\sqrt{2\pi}} e^{-y^2/2}.$$

In this case

$$T = \frac{(\bar{X} - \mu)}{S/\sqrt{n}}$$

is a pivotal quantity having the Student's  $t$ -distribution with  $n - 1$  degrees of freedom.

Let  $t_{p,n-1}$  be the  $(1 - p)$  quantile of Student's  $t$ -distribution with  $n - 1$  degrees of freedom. A  $(1 - \alpha)100\%$  confidence interval for  $\mu$  is

$$\left[ \bar{X} - t_{\alpha_2, n-1} \frac{S}{\sqrt{n}}, \bar{X} - t_{1-\alpha_1, n-1} \frac{S}{\sqrt{n}} \right],$$

where  $\alpha_1 + \alpha_2 = \alpha$ .

For any location-scale or scale family,  $S^2/\sigma^2$  is a pivotal quantity. This fact leads to confidence intervals for  $\sigma^2$  (or  $\sigma$ ).

When  $X_1, \dots, X_n$  are i.i.d.  $N(\mu, \sigma^2)$ ,

$$\frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2.$$

Let  $a$  and  $b$  be such that  $P(a < \chi_{n-1}^2 < b) = 1 - \alpha$ . Then

$$\left[ \frac{(n-1)S^2}{b}, \frac{(n-1)S^2}{a} \right]$$

is a  $(1 - \alpha)100\%$  confidence interval for  $\sigma^2$ .



## Shortest length confidence intervals

Subject to the confidence coefficient being  $1 - \alpha$ , we would like our confidence interval to be of shortest length.

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**Example 43 (continued)**  $X_1, \dots, X_n$  i.i.d.  $N(\mu, \sigma^2)$ . The length of the previously derived confidence interval for  $\mu$  is

$$\frac{S}{\sqrt{n}}(-t_{1-\alpha_1, n-1} + t_{\alpha_2, n-1}).$$

Subject to the constraint  $\alpha_1 + \alpha_2 = \alpha$ , choose  $\alpha_1$  and  $\alpha_2$  to minimize

$$t_{\alpha_2, n-1} - t_{1-\alpha_1, n-1}.$$

**Theorem 12** Let  $f$  be a unimodal pdf. If the interval  $[a, b]$  satisfies

(i)  $\int_a^b f(x) dx = 1 - \alpha$

(ii)  $f(a) = f(b) > 0$

(iii)  $a \leq x^* \leq b$  where  $x^*$  is the mode of  $f$ ,

then  $[a, b]$  is the shortest of all intervals that satisfy (i).

**Proof:** See Casella and Berger, p. 442.

Continuing Example 43, since the  $t$ -distribution is unimodal, we may apply Theorem 13. Now, the  $t$ -distribution is symmetric, so  $t_{\alpha_2, n-1} - t_{1-\alpha_1, n-1}$  is minimized subject to  $\alpha_1 + \alpha_2 = \alpha$  when  $\alpha_1 = \alpha_2 = \alpha/2$ .

The shortest interval has the form

$$\left[ \bar{X} - t_{\alpha/2, n-1} \frac{S}{\sqrt{n}}, \bar{X} + t_{\alpha/2, n-1} \frac{S}{\sqrt{n}} \right].$$

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In some situations we can minimize length directly.

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**Example 44** In Example 42 we had  $X_1, \dots, X_n$  i.i.d.  $U(0, \theta)$ , and found that  $[X_{(n)}/b, X_{(n)}/a]$  is a  $(1-\alpha)100\%$  confidence interval for  $\theta$  for each  $(a, b)$  such that  $b^n - a^n = 1 - \alpha$ . Here, we can't apply Theorem 12, but we could minimize the length directly.

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