

## Lezione 12 - Interval Estimation

The interval estimate is when we input the observed sample.

Confidence Interval for Gaussians with known  $\sigma^2$

$$\left( \bar{x}_n - z_{1-\frac{\alpha}{2}} \frac{\sigma_0}{\sqrt{n}}, \bar{x}_n + z_{1-\frac{\alpha}{2}} \frac{\sigma_0}{\sqrt{n}} \right)$$

The length is:  $L = 2 \cdot z_{1-\frac{\alpha}{2}} \sqrt{\frac{\sigma_0^2}{n}}$

For  $L \rightarrow 0 \Rightarrow n \rightarrow \infty$

If  $\alpha \rightarrow 0 \Rightarrow z_{1-\frac{\alpha}{2}} \rightarrow \infty \Rightarrow L \rightarrow \infty$

↳ it's not convenient to increase  $1 - \frac{\alpha}{2}$ , since it increases the length, so it is less precise.

### Example

A free fall falling 100 m.

$X$  = fall time (s)

$$n = 4$$

$$X_1, X_2, X_3, X_4 \stackrel{iid}{\sim} N(\mu, (0,8)^2)$$

$$x_1 = 4,4 \text{ s} \quad x_2 = 4,6 \text{ s} \quad x_3 = 4,7 \text{ s} \quad x_4 = 4,5$$

$$\hat{\mu} = \bar{x}_4 = 4,55$$

CI for  $\mu$  [two-sided] at level  $1 - \alpha = 95\%$

$$\left( \bar{x}_n - z_{1-\frac{0,95}{2}} \sqrt{\frac{(0,8)^2}{4}}, \bar{x}_n + z_{1-\frac{0,95}{2}} \sqrt{\frac{(0,08)^2}{4}} \right)$$

$$(3,766, 5,334) \rightarrow L = 1,568$$

Remark

↪ we can not say  $P(3,766 < \mu < 5,334) = 0,95$ ,  
 since there is no range variable involved, so  
 there is no probability as there is no randomness,  
 these are all observed and real values.

CI for  $\mu$  of confidence level  $1-\alpha$  is

$$\left( \underbrace{\bar{x}_n - z_{1-\frac{\alpha}{2}} \sqrt{\frac{\sigma_0^2}{n}}, \bar{x}_n + z_{1-\frac{\alpha}{2}} \sqrt{\frac{\sigma_0^2}{n}} }_{\text{K times}} \right)$$

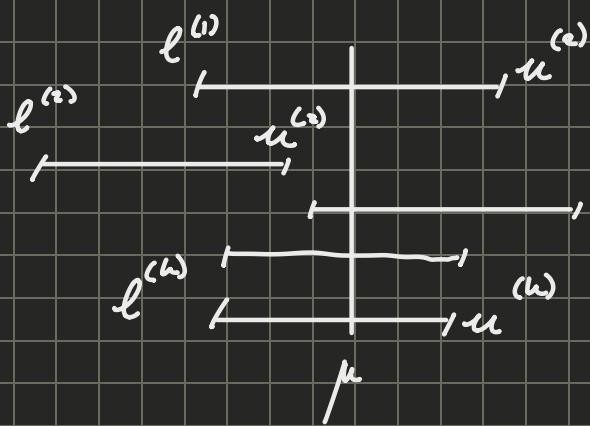
K → repeat the "experiment" K times, where K is large  
 e.g. K = 1000, or K = 10000 or more

$K=1$   $(x_1^{(1)}, x_2^{(1)}, \dots, x_n^{(1)})$  → first observed sample  
 we find  $l^{(1)}, u^{(1)}$

$K=2$   $(x_1^{(2)}, x_2^{(2)}, \dots, x_n^{(2)})$  → another real value interval  
 $(l^{(2)}, u^{(2)})$

$K=K=10000 \rightarrow$  → we get  $(l^{(u)}, u^{(u)})$

Suppose we know the true value of  $\mu$ :



Confidence level  $1-\alpha$ , means  $(1-\alpha) \cdot 100\%$  of  $k$  intervals contain the true value of  $\mu$ .

Definition:

A one-sided lower confidence bound (limit) for  $\theta$  of a confidence level  $100(1-\alpha)\%$  is determined by the random interval  $(L(X_1, \dots, X_n), \infty)$  such that:

$$P_{\theta} (L(X_1, \dots, X_n) < \theta) = 1-\alpha$$

There is a parallel for the upper limit.

What changes if the variance is unknown, while we are still trying to estimate  $\mu$ .

What changes is the distribution of the pivotal quantity, from  $N(0,1)$  to the t-student distribution.

Definition:

The random variable  $T$  is t-distributed with  $k$  degrees of freedom if  $T$  is absolutely continuous with density:

$$f_T(t) = \frac{\Gamma(\frac{k+1}{2})}{\sqrt{\pi k} \Gamma(\frac{k}{2})} \cdot \frac{1}{\left(1 + \frac{t^2}{k}\right)^{\frac{k+1}{2}}} \quad t \in \mathbb{R}$$

Prop 1: If  $k \geq 2$ ,  $E(T) = 0$

Prop 2: If  $k \geq 3$ ,  $\text{Var}(T) = \frac{k}{k-2}$

Prop 3: If  $k \rightarrow \infty$ :  $f_T(t) \rightarrow \phi(t)$   $\forall t$

$\hookrightarrow$  closeness of standard normal.

Notation:  $T \sim t(n)$ , quantile  $t_{1-\alpha}(n)$  (or  $t_{\alpha,n}$ )

$P(T \leq t_{1-\alpha}(n)) = 1 - \alpha \rightarrow$  same definition as  $t_{1-\alpha}$

T-student density plot.

$$P(T < t_{1-\alpha}(n)) > P(T < z_{1-\alpha})$$

$\hookrightarrow$  the tails are heavier, the have more mass.

CI ( $\mu$ ) of a Gaussian population with unknown variance

$X_1, \dots, X_n$  random sample from  $N(\mu, \sigma^2)$ ,  $\sigma^2$  unknown.

A two-sided confidence interval (CI) for  $\mu$  of confidence level

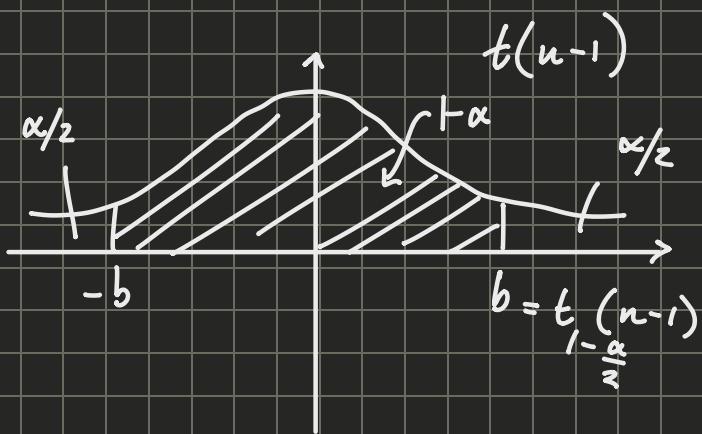
$1 - \alpha$  is:

$$\left( \bar{x}_n - \frac{s}{\sqrt{n}} t_{1-\frac{\alpha}{2}}(n-1), \bar{x}_n + \frac{s}{\sqrt{n}} t_{1-\frac{\alpha}{2}}(n-1) \right)$$

$$\hookrightarrow \text{estimate for } \sigma^2, \frac{1}{n-1} \sum_{j=1}^n (x_j - \bar{x}_n)^2$$

We get this from:

$$P\left(-b < \frac{\bar{X}_n - \mu}{\sqrt{\frac{s^2}{n}}} < b\right) = 1 - \alpha$$



$$P\left(-t_{1-\frac{\alpha}{2}}(n-1) < \frac{\bar{X}_n - \mu}{\sqrt{\frac{s^2}{n}}} < t_{1-\frac{\alpha}{2}}(n-1)\right) = 1 - \alpha$$

Same as before we isolate  $\mu$ :

$$\left(\bar{X}_n + t_{1-\frac{\alpha}{2}}(n-1)\right) \sqrt{\frac{s^2}{n}} < \mu < \bar{X}_n + t_{1-\frac{\alpha}{2}}(n-1) \sqrt{\frac{s^2}{n}}$$

CI for  $\mu$  of a population if  $n$  is large

In general if the are not gaussian we cannot find the CI for  $\mu$ , unless  $n$  is large ( $n \geq 50$ ). If this is true the distribution of the pivotal statistic can be assumed to be large  $N(0, 1)$

$$\frac{\bar{X}_n - \mu}{\sqrt{\frac{s^2}{n}}} \xrightarrow{\text{approx}} N(0, 1)$$

## Example

A company manufactures a new type of solar panels

$n = 128$  tested, only 9 out of 128 are found to be defective.

- a) Find a two-sided CI of 95% level for the probability that a new solar panel is defective.

$$n = 128$$

$$X_1, \dots, X_{128} \stackrel{iid}{\sim} \text{Be}(p)$$

$$X_i = \begin{cases} 1 & \text{if the } i\text{-th solar panel} \\ & \text{is defective} \\ 0 & \text{otherwise.} \end{cases}$$

$p$  = probability that a new solar panel is defective,  $p \in (0, 1)$

$$p = \mathbb{E}[X_i]$$

CI for the mean of an UNKNOWN population

Pivot

$$\frac{\bar{X}_n - p}{\sqrt{\frac{s^2}{n}}} \underset{n \text{ large}}{\underset{\text{approx}}{\sim}} N(0, 1)$$

$$P\left(-z_{1-\frac{\alpha}{2}} < \frac{\bar{X}_n - p}{\sqrt{\frac{s^2}{n}}} < z_{1-\frac{\alpha}{2}}\right) \approx 1 - \alpha$$

How can we approximate  $s^2$ ?

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

$$\begin{aligned}
 &= \frac{1}{n-1} \sum_{i=1}^n \left[ X_i^2 - 2\bar{X}_n X_i - \bar{X}_n^2 \right] \\
 &= \frac{1}{n-1} \left[ \left( \sum_{i=1}^n X_i^2 \right) - 2n\bar{X}^2 + n(\bar{X})^2 \right] \\
 &\boxed{\hat{\sigma}^2 = \frac{1}{n-1} \sum_i X_i^2 - n\bar{X}^2}
 \end{aligned}$$

In this case  $X_j = X_i$  since  $X_i$  is either 1 or 0.

$$\begin{aligned}
 &= \frac{1}{n-1} \left[ \sum_i X_i - n(\bar{X})^2 \right] = \\
 &= \frac{1}{n-1} \left[ n\bar{X} - n(\bar{X})^2 \right] = \underbrace{\frac{n}{n-1}}_{\approx 1} \bar{X}(1-\bar{X}) \approx \bar{X}(1-\bar{X})
 \end{aligned}$$

$$\Rightarrow \text{Pivot is } \frac{\bar{X} - p}{\sqrt{\frac{\bar{X}(1-\bar{X})}{n}}} \stackrel{\text{approx}}{\sim} N(0, 1)$$

CI of  $p$  with approximate confidence level  $1-\alpha$  is

$$\left( \bar{x}_n - z_{1-\frac{\alpha}{2}} \sqrt{\frac{\bar{x}_n(1-\bar{x}_n)}{n}}, \bar{x}_n + z_{1-\frac{\alpha}{2}} \sqrt{\frac{\bar{x}_n(1-\bar{x}_n)}{n}} \right)$$

$$\hat{p} = \bar{x}_n = \frac{9}{128} \rightarrow (0.072, 0.1146) = (7.6\%, 11.46\%)$$

↳ in formulation it might be written with  $\hat{p}$

$$1-\alpha = 95\% \quad z_{1-\frac{0.05}{2}} = z_{1-\frac{0.05}{2}} = 1.96$$

$$\begin{aligned}
 L &= 8.86\% \\
 &= 0.0886
 \end{aligned}$$

$$\begin{aligned}
 n\hat{p} &= n\bar{x}_n = 9 > 5 & n > 50 \\
 n(1-\hat{p}) &\approx 5
 \end{aligned}$$

b) Find the minimum number of solar panels such that the length of CI (with same  $1-\alpha$ ) is not longer than 5%.

$n$

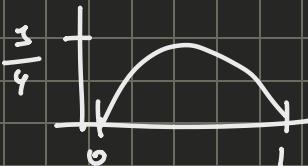
$$L = 2 z_{1-\frac{\alpha}{2}} \sqrt{\frac{\bar{x}_n(1-\bar{x}_n)}{n}} \leq 0,05$$

$$L = 2 z_{1-\frac{\alpha}{2}} \sqrt{\frac{\bar{x}_n(1-\bar{x}_n)}{n}} \leq 2 z_{1-\frac{\alpha}{2}} \sqrt{\frac{\frac{1}{4}}{n}} \leq 0,05 \Rightarrow \sqrt{n} \geq \frac{1,96}{0,05} \sqrt{\frac{15}{36}} = \frac{1,96}{0,05} \sqrt{15}$$

let's create the function

$$x \mapsto x(1-x) \leq \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$$

$$x \in (0,1)$$



$$\boxed{n \geq 1537}$$

since  $x$  includes  $n$ , we found a limit for it so we didn't have to solve weirdly.

Remark

CI for  $p$  approximate

$$\text{If } \hat{p} - z_{1-\frac{\alpha}{2}} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} < 0$$

$\bar{x}_n$

we do not unite  $CI(-\bar{x}_{min}, \bar{x}_{max})$

We have to unite  $CI(0, \hat{p} + z_{1-\frac{\alpha}{2}} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}})$   
we limit two

Similarly:

$$\text{If } \hat{p} + z_{1-\frac{\alpha}{2}} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} > 1 \quad CI \left( \hat{p} - z_{1-\frac{\alpha}{2}} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}, 1 \right)$$

We limit to 1.

C.I. for parameters different from  $\mu$ , using plug-in estimator

$$X_1, \dots, X_n \stackrel{iid}{\sim} \text{Exp}(\theta)$$

$$E_\theta[X_i] = \frac{1}{\theta} \quad \text{Var}(X_i) = \frac{1}{\theta^2} = E_\theta[(X_i)^2]$$

$$\bar{X}_n = \left( \frac{1}{\theta} \right) \quad \text{Plug-in estimator}$$

$$\text{Var}[\bar{X}_n] = (\bar{X}_n)^2$$

How do we find the confidence interval for

If  $n$  is large ( $n \geq 50$ ) a C.I. for  $\frac{1}{\theta}$  ( $= E_\theta[X_i]$ )

$$\left( \bar{X}_n - z_{1-\frac{\alpha}{2}} \sqrt{\frac{(\bar{X})^2}{n}}, \bar{X}_n + z_{1-\frac{\alpha}{2}} \frac{\bar{X}}{\sqrt{n}} \right)$$

$\hookrightarrow$  C.I. for  $\frac{1}{\theta}$

$$\left( \bar{X}_n - z_{1-\frac{\alpha}{2}} \frac{\bar{X}}{\sqrt{n}} < \frac{1}{\theta} < \bar{X}_n + z_{1-\frac{\alpha}{2}} \frac{\bar{X}}{\sqrt{n}} \right)$$

We can solve these inequalities with respect to  $\theta$

$\Rightarrow$  C.I. for  $\theta$  of approximate level  $1-\alpha$

$$\left( \frac{1}{\bar{X}_n - z_{1-\frac{\alpha}{2}} \frac{\bar{X}}{\sqrt{n}}}, \frac{1}{\bar{X}_n + z_{1-\frac{\alpha}{2}} \frac{\bar{X}}{\sqrt{n}}} \right)$$