

## CONFIDENCE INTERVALS

The objective is the construction of confidence intervals for population parameters. When the underlying distribution of a statistic is approximately normal <sup>1</sup>, the confidence interval will be of the form

$$\text{Point estimate} \pm \text{margin of error}$$

The question becomes: how do we build the margin of error, i.e. how to identify the appropriate test statistics.

### —→ **When $\sigma$ is known**

If the population standard deviation  $\sigma$  is known, a  $(100-\alpha)\%$  confidence interval for  $\mu$  is given by:

$$\bar{X} \pm z_{\alpha/2} \times \frac{\sigma}{\sqrt{n}}$$

$\bar{X}$ : sample mean;  $n$ : sample size;  $\sigma$ : population standard deviation

$z$ : z value from the z-test statistics;  $\alpha$ : significance level

$z_{\alpha/2} = 1.96$  for a 95% confidence interval

Use cases:

- estimation of a population proportion  $p$  (how close is the sample proportion  $\hat{p}$  to the true value of the parameter)
- estimation of a population mean in the unlikely case that  $\sigma$  is known

### —→ **When $\sigma$ is unknown**

If the population standard deviation  $\sigma$  is not known, a  $(100-\alpha)\%$  confidence interval for  $\mu$  is given by:

$$\bar{X} \pm t_{\alpha/2} \times \frac{s}{\sqrt{n}}$$

We use  $s$  the unbiased sample standard deviation to estimate  $\sigma$  and therefore need to implement the t-procedure instead of the z procedure.

$\bar{X}$ : sample mean;  $n$ : sample size;  $s$ : unbiased sample standard deviation

$t$ : t value from the t distribution with  $n - 1$  degrees of freedom

$\alpha$ : significance level

---

<sup>1</sup>One of the major justification for supposing a normal distribution is the central limit theorem.