

Design and Analysis of Experiments

04 - Statistical Intervals

Version 2.11

Felipe Campelo

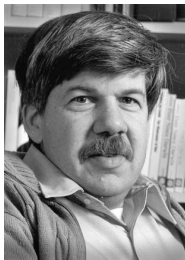
<http://www.cpdee.ufmg.br/~fcampelo>

Graduate Program in Electrical Engineering

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“Science is an integral part of culture. It’s not this foreign thing, done by an arcane priesthood. It’s one of the glories of the human intellectual tradition.”

Stephen Jay Gould
1941-2002
American paleontologist



Statistical Intervals

Introduction

Statistical intervals are important in quantifying the uncertainty associated to a given estimate;

As an example, let's recap the coaxial cables example: *a coaxial cable manufacturing operation produces cables with a target resistance of 50Ω and a standard deviation of 2Ω . Assume that the resistance values can be well modeled by a normal distribution.*

Let us now suppose that a sample mean of $N = 25$ observations of resistance yields $\bar{x} = 48$. Given the sampling variability, it is very likely that this value is not exactly the true value of μ , but we are so far unable quantify how much uncertainty there is in this estimate.

Statistical Intervals

Definition

Statistical intervals define regions that are likely to contain the true value of an estimated parameter.

More formally, it is generally possible to quantify the level of uncertainty associated with the estimation, thereby allowing the derivation of sound conclusions at predefined levels of certainty.

Three of the most common types of interval are:

- Confidence intervals;
- Tolerance intervals;
- Prediction intervals;

Confidence Intervals

Definition

Confidence intervals quantify the degree of uncertainty associated with the estimation of population parameters such as the mean or the variance.

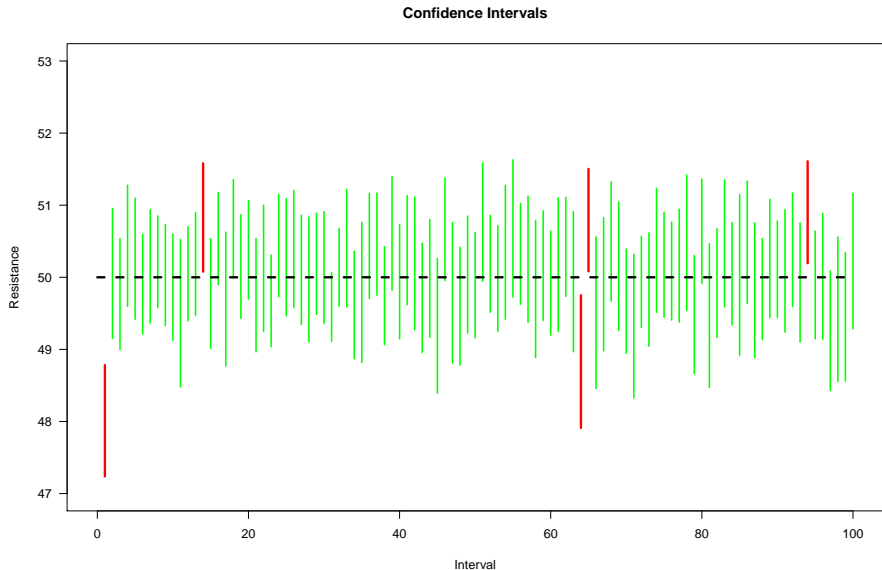
Can be defined as “*the interval that contains the true value of a given population parameter with a confidence level of $100(1 - \alpha)$* ”;

- **Wrong:** “there is a 95% chance that the interval contains the true population mean.”
- **right:** “The method used to derive the interval has a hit rate of 95%” - i.e., the interval generated has a 95% chance of “capturing” the true population parameter.

Easier to understand if you think about the confidence level as a confidence in the **method**, not in the interval.

Confidence Intervals

Example: 100 $CI_{.95}$ for a sample of 25 observations



Confidence Intervals

CI on the Mean of a Normal Variable

The two-sided $CI_{(1-\alpha)}$ for the mean of a normal population with known variance σ^2 is given by:

$$\bar{y} - \frac{\sigma}{\sqrt{N}} z_{(\alpha/2)} \leq \mu \leq \bar{y} + \frac{\sigma}{\sqrt{N}} z_{(\alpha/2)}$$

wherein $(1 - \alpha)$ is the confidence level and $z_{(\alpha/2)}$ is the $(1 - \alpha/2)$ -quantile of the standard normal distribution.

For the more usual case with an unknown variance,

$$\bar{y} - \frac{s}{\sqrt{N}} t_{(\alpha/2; N-1)} \leq \mu \leq \bar{y} + \frac{s}{\sqrt{N}} t_{(\alpha/2; N-1)}$$

wherein $t_{(\alpha/2; N-1)}$ is the corresponding quantile of the t distribution with $N - 1$ degrees of freedom.

Confidence Intervals

CI on the Variance of a Normal Variable

Similarly, a two-sided confidence interval on the variance of a normal variable can be easily calculated:

$$\frac{(N-1)s^2}{\chi_{\alpha/2; N-1}^2} \leq \sigma^2 \leq \frac{(N-1)s^2}{\chi_{1-\alpha/2; N-1}^2}$$

wherein $\chi_{\alpha/2; N-1}^2$ and $\chi_{1-\alpha/2; N-1}^2$ are the upper and lower $(\alpha/2)$ -quantiles of the χ^2 distribution with $N-1$ degrees of freedom.

Tolerance Intervals

Definition

*“A tolerance interval is an **enclosure** interval for a specified proportion of the sampled population, not its mean or standard deviation. For a specified confidence level, you may want to determine lower and upper bounds such that 99 percent of the population is contained within them.”^[1].*

[1] J.G. Ramírez: https://www.sas.com/resources/whitepaper/wp_4430.pdf

Bibliography

Required reading

- 1 J.G. Ramírez, *Statistical Intervals: Confidence, Prediction, Enclosure*:
https://www.sas.com/resources/whitepaper/wp_4430.pdf
- 2 D.C. Montgomery and G.C. Runger, *Applied Statistics and Probability for Engineers*, Chapter 8. 3rd Ed., Wiley 2005.

Recommended reading

- 1 Simply Statistics (blog) - <http://simplystatistics.org>
- 2 R. Dawkins, *Climbing Mount Improbable*, W.W.Norton&Co.,1997.

About this material

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