M5S3 - Interpretation of Confidence Intervals

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Outline

- Interpretation of probability
 - Frequentist
 - Bayesian

Frequentist interpretation of probability

Interpretation

The frequentist interpretation of probability is that probability is the long-run relative frequency of an event.

Thus, if we have a sequence of independent and identically distributed binary random variables I_1, I_2, \ldots where I_i is the indicator of the event occurring in the ith trial, i.e.

$$I_i = \left\{ egin{array}{ll} 1 & \mbox{if the event occurs in the ith trial} \\ 0 & \mbox{if the event does not occur in the ith trial}, \end{array} \right.$$

the probability p is defined as

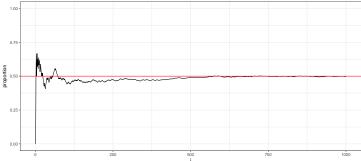
$$p = \lim_{n \to \infty} \frac{\sum_{i=1}^{n} I_i}{n}.$$

Coin flipping example

Let I_i be the indicator that the ith coin flip is heads, i.e.

$$I_i = \left\{ \begin{array}{l} 1 & \text{if he ith coin flip is heads} \\ 0 & \text{if the ith coin flip is not heads,} \end{array} \right.$$

Now we define the probability as the proportion of heads as the number of flips tends to infinity.

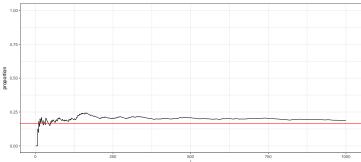


Die rolling example

Let I_i be the indicator of the *i*th die roll being a 1, i.e.

$$I_i = \left\{ egin{array}{ll} 1 & \mbox{if the ith die roll is 1} \\ 0 & \mbox{if the ith die roll is not 1}, \end{array} \right.$$

Now we define the probability as the proportion of 1s as the number of rolls tends to infinity.



Construction of confidence intervals

Recall that the formula for a $100(1-\alpha)\%$ confidence interval (CI) based on the standard normal is

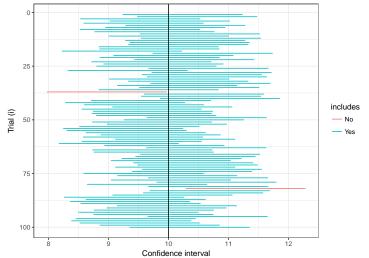
$$\overline{X} \pm z_{\alpha/2} \frac{\sigma}{\sqrt{n}}.$$

We obtained this interval by calculating the following probability

$$P\left(\overline{X} - z_{\alpha/2} \frac{\sigma}{\sqrt{n}} < \mu < \overline{X} + z_{\alpha/2} \frac{\sigma}{\sqrt{n}}\right) = 1 - \alpha.$$

Thus a confidence interval has random endpoints since \overline{X} is random. We can imagine performing this procedure repeatedly and calculating the proportion of times the CI includes μ .

Let $X_i \stackrel{iid}{\sim} N(10,1^2)$ with n=4. Then a 95% CI based on the Empirical Rule is $\overline{X} \pm 2 \cdot 1/\sqrt{4} = \overline{X} \pm 1$.



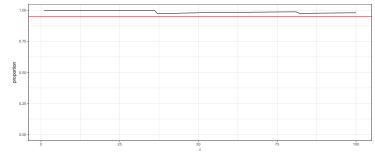
Interpretation of Confidence Intervals

Let I_i be the indicator that the ith $100(1-\alpha)\%$ confidence interval (CI) for the population mean μ contains μ , i.e.

$$I_i = \left\{ \begin{array}{ll} 1 & \text{if the ith CI includes μ} \\ 0 & \text{if the ith CI does not include μ,} \end{array} \right.$$

Then

$$\lim_{n \to \infty} \frac{\sum_{i=1}^{n} I_i}{n} = 1 - \alpha.$$



Relation to binomial distribution

Recall that a random variable Y has a binomial distribution if

$$Y = \sum_{i=1}^{n} I_i$$

where I_i are independent and identically distributed (iid) Bernoulli random variables with a common probability of success p. Here

$$I_i = \begin{cases} 1 & \text{if Cl } i \text{ includes } \mu \\ 0 & \text{if Cl } i \text{ does not include } \mu \end{cases}$$

Since each CI has probability $1-\alpha$ of including μ , we have $I_i \stackrel{iid}{\sim} Ber(1-\alpha)$ and $Y \sim Bin(n, 1-\alpha)$.

Expected number of CIs that cover μ

If we construct n Cls each with probability $1-\alpha$ of including μ , then how many CIs do we expect will include μ . Since Y is the random number of Cls that will include the truth and $Y \sim Bin(n, 1 - \alpha)$, we have

$$E[Y] = n(1 - \alpha).$$

Calculate the expected number that will include the truth for the following scenarios:

- $n = 100, 1 \alpha = 0.95$ then $E[Y] = 100 \cdot 0.95 = 95$
- $n = 1000, 1 \alpha = 0.7$ then $E[Y] = 1000 \cdot 0.70 = 700$
- $n = 77, 1 \alpha = 0.66$ then $E[Y] = 77 \cdot 0.66 = 50.82$

If we are interested in how many will not cover the truth, this is the random variable n-Y and E[n-Y]=n-E[Y]. Calculate the expected number that will not include the truth for the same scenarios:

- $n = 100, 1 \alpha = 0.95$ then E[n Y] = 100 95 = 5
- $n = 1000, 1 \alpha = 0.7$ then E[n Y] = 1000 700 = 300
- $n = 77, 1 \alpha = 0.66$ then E[n Y] = 77 50.82 = 26.18

Summary

Here are the interpretation statements for a $100(1-\alpha)\%$ confidence interval for the population mean μ :

- Out of $n \ 100(1-\alpha)\%$ confidence interval for μ , we expect $n(1-\alpha)$ confidence intervals to include/cover μ (and $n\alpha$ to not cover μ .
- We are $100(1-\alpha)\%$ confident that μ falls within the bounds of the constructed interval.

I really hate the second statement as I believe it gives you a false impression of what you have actually learned. The second interpretation DOES NOT tell you what you should believe, it is really a succinct version of the prevous interpretation.

When you see the words confidence or confident, think in your head, the word frequency.

Issues with a frequentist interpretation of probability

How can you interpret the following probability statements:

- What is the probability it will rain tomorrow?
- What is the probability the Vikings will their next game?
- What is the probability my unborn child has Down syndrome?
- What is the probability humans are the main cause of climate change on Earth?

Bayesian interpretation of probability

Interpretation

The Bayesian interpretation of probability is that probability is a statement about your degree of belief that an event will (or has) occurred.

Advantages:

- Can interpret probability for one time events.
- States what you should believe.
- Natural to make decisions based on your belief.
- Everyone has their own probability.

Disadvantages:

- Requires more math (integration).
- Requires you to specify your belief before seeing data.
- Has no relation to relative frequency.
- Everyone has their own probability.

Credible intervals

The Bayesian analog to confidence intervals are credible intevals. These intervals tell you where you should believe the parameter to be. Thus a $100(1-\alpha)\%$ credible interval for μ tells you that you should believe that the true population mean μ is in the interval with probability $1-\alpha$.

It turns out, under a particular prior, the confidence intervals that we construct are exactly the same as credible intervals. Thus, you will actually be correct even when you misinterpret confidence intervals.