

An example of interval estimation

Problem 47 from Section 6.7 of Durrett's book:

47. Of the first 10,000 votes cast in an election, 5,180 were for candidate A. Find a 95% confidence interval for the fraction of votes that candidate A will receive.

Solution. We try to mimic the reasoning in the voltage measurement example:

1. We would like to estimate the fraction p of votes for candidate A among all the votes cast in the election. Whether the vote of a randomly picked person is for candidate A or not can be modelled with a Bernoulli random variable X with parameter p (why?).
2. The available statistical evidence is the votes of 10,000 individuals, that is, a random sample of size 10,000 from the entire population of those who cast votes. This can be modelled as a collection

$$X_1, X_2, \dots, X_{10000}$$

of 10,000 i.i.d. Bernoulli random variables each with parameter p .

3. A reasonable point estimator for p is the fraction

$$\hat{p} = \frac{X_1 + X_2 + \dots + X_{10000}}{10000}$$

of the sampled individuals who vote for candidate A. In our model, this is a random variable. Its observed value is $5180/10000 = 0.5180$

4. What can we say about the distribution of $\hat{p}$?

- Since $\hat{p}$ is the average of a relatively large number of i.i.d. random variables, by the CLT, the distribution of $\hat{p}$ is approximately normal.
- The parameters of the approximating normal distribution are the mean and variance of $\hat{p}$, that is,

$$\begin{aligned} \mu = \mathbb{E}[\hat{p}] &= \mathbb{E} \left[\frac{X_1 + X_2 + \dots + X_{10000}}{10000} \right] = p, \\ \sigma^2 = \mathbb{V}\text{ar}[\hat{p}] &= \mathbb{V}\text{ar} \left[\frac{X_1 + X_2 + \dots + X_{10000}}{10000} \right] = \frac{p(1-p)}{10000}. \end{aligned}$$

(Here, we have used the fact that the variance of a Bernoulli random variable with parameter p is $p(1-p)$.)

- Alternatively, we can observe that

$$N = 10000\hat{p} = X_1 + X_2 + \dots + X_{10000}$$

is a binomial random variable with parameters $n = 10000$ and (unknown) p .

5. We would like to use $\hat{p}$ to construct an interval estimator for p with a confidence level of 95%. The interval estimator will (often) be of the form $\hat{p} \pm \varepsilon$. We have to choose ε such that $\mathbb{P}(p = \hat{p} \pm \varepsilon) \approx 95\%$.

Since $\hat{p}$ is approximately distributed according to $N(p, \frac{p(1-p)}{10000})$, its normalized version

$$\frac{\hat{p} - p}{\sqrt{p(1-p)/10000}} = \frac{\hat{p} - p}{\sqrt{p(1-p)/100}}$$

is approximately $N(0, 1)$. Hence, for every $a > 0$, we have

$$\mathbb{P} \left(-a < \frac{\hat{p} - p}{\sqrt{p(1-p)/100}} < a \right) \approx \Phi(a) - \Phi(-a) = 2\Phi(a) - 1,$$

where Φ is the cdf of the standard normal distribution. Setting $2\Phi(a) - 1 = 95\%$, we get $\Phi(a) = 0.975$, so we can use a computer (or the App for the normal distribution) to find $a \approx 1.95996 \approx 1.96$. Thus,

$$\mathbb{P}\left(-1.96 < \frac{\hat{p} - p}{\sqrt{p(1-p)/100}} < 1.96\right) \approx 95\% .$$

Here, there is a slight complication: if we naively proceed as in the measurement example, we obtain

$$\mathbb{P}\left(p = \hat{p} \pm 1.96 \times \sqrt{p(1-p)/100}\right) = 95\% ,$$

but the interval

$$\hat{p} \pm 1.96 \times \sqrt{p(1-p)/100} \quad (\clubsuit)$$

depends on the parameter p that we are trying to estimate!

To resolve this issue, we can follow at least three different approaches:

Approach 1: (rough)

In $\sqrt{p(1-p)/100}$, quietly replace p with $\hat{p}$ and pretend nobody notices. This way, we obtain the interval estimator

$$\hat{p} \pm 1.96 \times \sqrt{\hat{p}(1-\hat{p})/100}$$

Since the sample size is large, $\hat{p}$ is expected to be close to p , so the answer we obtain using this approach should not be too off the mark. In other words, the confidence level of the above estimator should be close to 95%.

Approach 2: (conservative but rigorous)

Note that $p(1-p)$ is maximized when $p = 1/2$ (why?). Hence, the interval $(\clubsuit)$ is included in the interval

$$\hat{p} \pm 1.96 \times \sqrt{1/2(1-1/2)/100} = \hat{p} \pm 0.0098 .$$

The confidence level of this interval estimator will be larger than 95%.

Approach 3: (solving the quadratic inequality)

The event $p = \hat{p} \pm 1.96 \times \sqrt{p(1-p)/100}$ can equivalently be written as

$$p - \hat{p} = \pm 1.96 \times \sqrt{p(1-p)/100}$$

or equivalently,

$$(p - \hat{p})^2 < 1.96^2 \times \frac{p(1-p)}{10000} .$$

The latter can in turn be written in the standard form $Ap^2 + Bp + C < 0$ for some A, B, C . (Here, A is a constant, while B and C depend on $\hat{p}$.) Solving this inequality for p , we obtain $P_1 < p < P_2$ where P_1 and P_2 are the roots of the quadratic polynomial $Ap^2 + Bp + C$. Thus, we obtain an interval estimator (P_1, P_2) with (almost accurate) confidence level 95%.

6. Plugging in the observed value 0.5180 for $\hat{p}$ in the interval estimator, we obtain an interval estimate for p .

<u>Approach 1:</u>	$p = 0.5180 \pm 0.009794$	(with roughly 95% confidence)
<u>Approach 2:</u>	$p = 0.5180 \pm 0.0098$	(with at least 95% confidence)
<u>Approach 3:</u>	$p = 0.5179931 \pm 0.009791769$	(with 95% confidence)

(In the third approach, the quadratic polynomial has the roots 0.5082013 and 0.5277849. We have written the interval $(0.5082013, 0.5277849)$ in the form $0.5179931 \pm 0.009791769$ for the sake of comparison with the other two approaches.)

7. The interpretation of the 95% confidence level is as follows: If we take many many similar random samples of size 10,000 from the voters, and each time use the same interval estimator to form a confidence interval, then for about 95% of such samples, the obtained interval will contain the true value of the fraction p .

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