Lecture 22 Confidence Interval For The Proportion

BIO210 Biostatistics

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Credit: "Ideas on protein synthesis (Oct. 1956)". Wellcome Collection.

Sample Proportion Example

Gene expression (over-simplified RNA-seq): We know the probability of detecting *PLK1* is $\pi = 0.000001126088083$. If we take a random sample of $n = 1,000,000$ mRNA molecules, what is the sampling distribution of proportion of *PLK1*?

$$
B(n,p)
$$
\n
$$
\begin{cases}\n\therefore \mathcal{N}(\mu = np, \sigma^2 = npq) & , \text{ when } np \geqslant 10 \text{ and } nq \geqslant 10 \\
\therefore Pois(\lambda = np) & , \text{ when } n \text{ is large, and } p \text{ is small,} \\
\text{ such that } np \text{ is between 0 and 10.} \\
\sim B(n,p) & , \text{ otherwise}\n\end{cases}
$$

The Limitations on *np* and *nq*

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- *•* Binomial: all data are within [0*, n*]
- Normal: no bounds $(-\infty, +\infty)$ for data, but most are within $[\mu 3\sigma, \mu + 3\sigma]$
- Intuitively: when $[\mu 3\sigma, \ \mu + 3\sigma]$ is within $[0, n]$, the approximation works well!

$$
\mu - 3\sigma > 0
$$
\n
$$
np - 3\sqrt{npq} > 0
$$
\n
$$
np > 3\sqrt{npq}
$$
\n
$$
np > 3\sqrt{npq}
$$
\n
$$
np > 9q
$$
\n
$$
np > 9(1-p) = 9 - 9p
$$
\n
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np > 9(1-q) = 9 - 9q
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\n
$$
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$$
\n
$$
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$$

Goal: for a population containing an unknown proportion (π) of data of our interest, find *a* and *b*, such that $\mathbb{P} (a \leq \pi \leq b) = 0.95$.

$$
\mathbb{P}(-1.96 \leq Z \leq 1.96) = 0.95
$$

$$
\mathbb{P}\left(-1.96 \leq \frac{p-\mu_P}{\sigma_P} \leq 1.96\right) = 0.95
$$

$$
\mathbb{P}\left(-1.96 \leq \frac{p-\pi}{\sqrt{\frac{\pi(1-\pi)}{n}}} \leq 1.96\right) = 0.95
$$

$$
\mathbb{P}\left(p-1.96\sqrt{\frac{\pi(1-\pi)}{n}} \leq \pi \leq p+1.96\sqrt{\frac{\pi(1-\pi)}{n}}\right) = 0.95
$$

95% CI For The Sample Proportion

The Wald Interval:

$$
\[p-1.96\sqrt{\frac{p(1-p)}{n}}, \, p+1.96\sqrt{\frac{p(1-p)}{n}}\,\right]\]
$$

• **Not using** *t*-distribution? - You don't need to! Remember $\sigma_{P} =$ $\sqrt{\pi(1-\pi)}$ $\frac{1}{n}$, and when *p* is calculated to estimate π , then σ_{P} is automatically determined, unlike in the situation of the mean, where you have to do extra (independent) calculation of s to estimate σ , which causes the extra error.

Probability vs. Statistics

- Probability: Previous studies showed that the drug was 80% effective. Then we can anticipate that for a study on 100 patients, on average 80 will be cured and at least 65 will be cured with 99.99% chance.
- Statistics: We observe that 78/100 patients were cured by the drug. We will be able to conclude that we are 95% confident that for other studies the drug will be effective on between 69.88% and 86.11% of patients.

Estimate Sample Size: We want to estimate the percentage of people cured by the drug. Suppose we could draw a truly random sample, and we want a 95% confidence interval estimation with a margin of error no more than *±* 2%. What is the smallest sample size required to obtain the desired margin of error ?

95% confidence interval:
$$
p \pm 1.96\sqrt{\frac{p(1-p)}{n}}
$$

Goal: find the smallest n such that it guarantees that $1.96\,$ $\sqrt{p(1-p)}$

$$
11/13
$$

 $\frac{1}{n} \leqslant 0.02$

- 1. Random Samples
- 2. Normal Condition: the sampling distribution of *p* needs to be normal
	- $np \geqslant 10$
	- $nq \geqslant 10$
- 3. Independence $(n < 10\%$ population size)
- Wilson score interval
- *•* Jeffreys interval
- *•* Agresti–Coull interval
- *•* Arcsine transformation
- *•* Clopper–Pearson interval (the exact method)