
Friday, March 19th 2021

Wk 7, Fr

Topic:: Central Limit Theorem

Central Limit Theorem

In summary, here is the take-away from the **Central Limit Theorem**.

Suppose you have a random sample of size n that is either

- i.i.d., or (like sampling w/ replacement)
- an SRS, with the sample size n being no more than 10% of the size of the population.

In the case that

- $n \geq 30$ {
1. the variable under consideration is quantitative, having population mean μ and standard deviation σ , then the sampling distribution for the sample mean \bar{X} is approximately $\text{Norm}(\mu, \sigma / \sqrt{n})$ for n large enough.

center of sampling dist for \bar{X} : μ ; $SE_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$

2. the variable under consideration is binary categorical, having population proportion p , then the sampling distribution for the sample proportion \hat{p} is approximately $\text{Norm}(p, \sqrt{p(1-p)/n})$ for n large enough.

Rules of thumb: n is large enough when

$$SE_{\hat{p}} = \sqrt{\frac{p(1-p)}{n}}$$

$$np \geq 10$$

$$n(1-p) \geq 10$$

Since

- null distributions
- randomization distributions
- bootstrap distributions

are all specialized versions of sampling distributions, then so long as the sample statistic in question is the sample's *mean* \bar{X} or the sample *proportion* \widehat{p} , we can expect the CLT to apply to these as well.

Explorations using apps at

https://onlinestatbook.com/stat_sim/sampling_dist/index.html

<https://shiny.calvin.edu:3838/scofield/samplingDists/>

<https://shiny.calvin.edu:3838/scofield/cltProportions/>