

7.1. Basic Properties of CI

- Example 7.1

Sample $n=31$. $\bar{x} = 80$. Normal Distribution.
with $\sigma = 2$. Q: CI of μ ?

Since $n=31 > 30$, sample mean. \bar{x} is normally distributed
with $E=\mu$ and $SD = \frac{\sigma}{\sqrt{n}}$

$$\text{Thus } Z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}}$$

$$P(-1.86 < Z < 1.86) = 0.95.$$

Thus, the interval $\bar{x} \pm 1.86 \cdot \frac{\sigma}{\sqrt{n}}$ or $(\bar{x} - 1.86 \cdot \frac{\sigma}{\sqrt{n}}, \bar{x} + 1.86 \cdot \frac{\sigma}{\sqrt{n}})$

- Define: 95% of CI for μ .

$$(\bar{x} - 1.86 \cdot \frac{\sigma}{\sqrt{n}}, \bar{x} + 1.86 \cdot \frac{\sigma}{\sqrt{n}})$$

*observed population
sample mean SD.*

- Other levels of Confidence Interval.

A 100(1- α)% confidence Interval:

$$\bar{x} \pm z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}$$

Example 7.3: 90% of CI

$$\Rightarrow \alpha = 0.1, z_{\alpha/2} = z_{0.05}$$

\Rightarrow Find 0.95% on z-table

$$\Rightarrow 1.645$$

- Confidence Level, Precision and Sample Size.
 - The price for higher CI is wider Interval.
 - Sample 7.4.

Normally distributed. $\sigma = 2$.

Q: What sample size n must satisfy 95% of CI has a width of at most 10.

$$10 = 2 \cdot (1.96) \left(\frac{\sigma}{\sqrt{n}} \right)$$

$$W = 2 \times Z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}$$

$$Z_{\alpha/2} \frac{\sigma}{\sqrt{n}} \rightarrow 10 \rightarrow$$

- Define:

The sample size necessary for the CI is

$$n = \left(2 \cdot Z_{\alpha/2} \cdot \frac{\sigma}{W} \right)^2$$

7.2 Large sample Confidence interval for a population mean and proportion.

- Proposition: $\rightarrow n > 40$.

If n is sufficiently large, by CLT, \bar{x} has approximately normal distribution. Therefore, large-sample CI for μ with CI approximately 100(1- α)% is $\bar{x} \pm Z_{\alpha/2} \frac{S}{\sqrt{n}}$ → replace population σ with S .

- A confidence interval for a population proportion.

- Proposition:

$$\text{Let } \hat{p} = \frac{\hat{p} + Z_{\alpha/2}^2 / 2n}{1 + Z_{\alpha/2}^2 / n}$$

Then a confidence interval for a population proportion p with confidence level approximately 100(1- α)% is.

$$\hat{p} \pm Z_{\alpha/2} \frac{\sqrt{\hat{p} \hat{q} / n + Z_{\alpha/2}^2 / 4n^2}}{1 + Z_{\alpha/2}^2 / n}$$

$$\hat{q} = 1 - \hat{p}$$

- One-Sided Confidence intervals.
 - Proposition:
 - A large-sample upper confidence bound for μ is.
- $\mu < \bar{x} + \frac{z_{\alpha}}{\sqrt{n}}$. Not $z_{\alpha/2}$.
- lower
- $\mu > \bar{x} - \frac{z_{\alpha}}{\sqrt{n}}$.
- General large-sample CI.
- $\hat{\theta}$ from sample $\hat{\theta}$ estimator.
- $$P\left(-z_{\alpha/2} < \frac{\hat{\theta} - \theta}{\sigma_{\hat{\theta}}} < z_{\alpha/2}\right) \approx 1 - \alpha.$$

7.3. Intervals based on a Normal Population Distribution

- When n is small, CLT cannot be invoked.
so T -distribution is introduced.

Theorem:

When \bar{x} is the mean of a random sample of size n from a normal distribution with mean μ .

$$T = \frac{\bar{x} - \mu}{S/\sqrt{n}} \quad (\text{t-distribution})$$

with $n-1$ degree of freedom. (v).

- Properties of T -distribution

- Each t -distribution is more spread out than standard normal curve.

- As v increase, the spread increase.

- One-sample t CI
 - $P(C - t_{\alpha/2, n-1} < T < t_{\alpha/2, n-1}) = 1 - \alpha$
 - $100(1-\alpha)\%$ CI for μ :

$$(\bar{x} - t_{\alpha/2, n-1} \cdot \frac{s}{\sqrt{n}}, \bar{x} + t_{\alpha/2, n-1} \cdot \frac{s}{\sqrt{n}}).$$
 - One-sided CI

$$\bar{x} \pm t_{\alpha, n-1} \cdot \frac{s}{\sqrt{n}}.$$

- Prediction Interval for a single future value.
 - Define: A Prediction Interval (PI) for a single observation to be selected from future interval:

$$\bar{x} \pm t_{d/2, n-1} \cdot S \sqrt{1 + \frac{1}{n}}.$$

- Tolerance Interval.
 - Let k be a number between 0 and 100. A tolerance interval for capturing at least $k\%$ of the values in a normal distribution with confidence level of $1 - \alpha$.

$$\bar{x} \pm \text{tolerance value} \cdot S.$$