

# Chapter 9

## Interval Estimation

### Detailed Solutions

#### 9.1 Basic Concept

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##### 9.1 True or False: The Meaning of Confidence

**Problem:** 95% CI is  $[10, 20]$ . Evaluate statements.

**Solution:**

- (a) **False.** The population mean  $\mu$  is a fixed constant, not a random variable. It is either in the interval or it is not. We cannot assign a probability to a fixed event after the fact.
- (b) **True.** This is the correct frequentist interpretation. "Confidence" refers to the reliability of the method (the interval construction process), not the specific result of one experiment.
- (c) **False.** A Confidence Interval estimates the *population parameter* (mean), not the distribution of individual data points. (That would be a Tolerance Interval).
- (d) **True.** Before data collection, the interval boundaries  $(\bar{X} \pm E)$  are random variables. After collection, they become fixed numbers  $[10, 20]$ .

(a) False    (b) True    (c) False    (d) True

#### 9.2 Factors Affecting Width

**Problem:** Width  $W = 2 \times z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$ .

**Solution:**

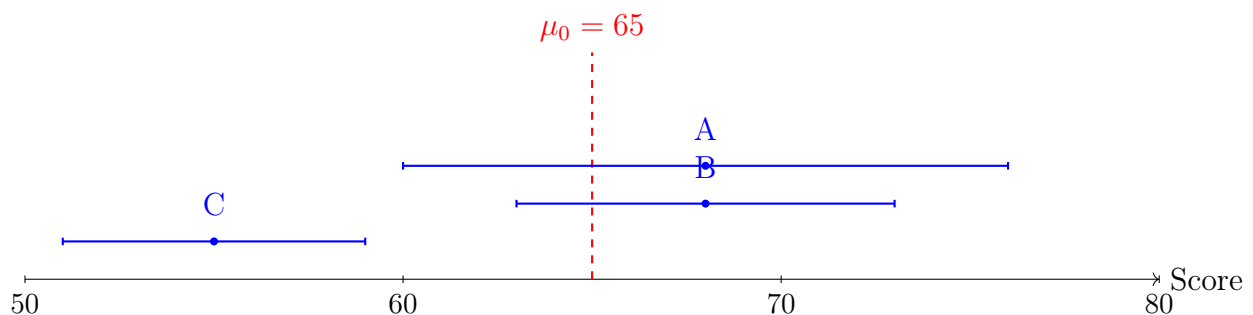
- (a) **Increase  $n$  (10 to 100): Narrower.**  $n$  is in the denominator ( $\sqrt{n}$ ). More data reduces uncertainty.

- (b) **Increase Confidence (90% to 99%): Wider.** To be more confident that we capture the true mean, we must cast a wider net ( $z$ -score increases).
- (c) **Larger  $s$ : Wider.** More variability in the data (noise) makes precise estimation harder.

(a) Narrower    (b) Wider    (c) Wider

### 9.3 Visual Forensics

**Problem:** Interpreting CIs from the chart.



**Solution:**

- (a) **Sample Size:** Study B likely used a **larger sample size**. Both have the same center and assumed same variance, but Interval B ( $73 - 63 = 10$ ) is narrower than Interval A ( $76 - 60 = 16$ ). Since  $\text{Width} \propto 1/\sqrt{n}$ , a narrower width implies larger  $n$ .
- (b) **Significance ( $\mu_0 = 70$ ):** Interval A is  $[60, 76]$ . Since 70 is **inside** the interval, the mean is **not significantly different** from 70 at the 5% level. We cannot reject the null hypothesis that  $\mu = 70$ .
- (c) **Overlap:** Interval B  $[63, 73]$  and Interval C  $[51, 59]$  do **not overlap**. This strongly suggests that the population means are significantly different.
- (d) **Precision:** Study C has the narrowest interval ( $59 - 51 = 8$ ). Thus, Study C provides the most precise estimate.

(a) Study B (Narrower)  
 (b) Not significant (70 is inside)  
 (c) No overlap  
 (d) Study C (Narrowest width)

## 9.2 Intermediate

### 9.4 CI for Mean (Raw Data)

**Problem:** Data: {52, 48, 56, 45, 50, 53, 49, 51, 54, 42}.  $n = 10$ .

**Solution:**

(a) **Stats:**  $\sum x = 500 \implies \bar{x} = 50.0$ .  $s^2 = \frac{\sum(x_i - \bar{x})^2}{n-1} = \frac{148}{9} \approx 16.44$ .  $s = \sqrt{16.44} \approx 4.055$ .

(b) **95% CI:** Since  $n < 30$  and  $\sigma$  unknown, use t-distribution with  $df = 9$ .  $t_{0.025,9} = 2.262$ .

$$\begin{aligned}\bar{x} \pm t \frac{s}{\sqrt{n}} &= 50.0 \pm 2.262 \frac{4.055}{\sqrt{10}} \\ &= 50.0 \pm 2.262(1.282) \\ &= 50.0 \pm 2.90.\end{aligned}$$

$$\text{CI} = [47.1, 52.9] \text{ MPa}.$$

- (c) **Interpretation:** The standard requires  $\mu \geq 48$ . The interval is  $[47.1, 52.9]$ . Since the lower bound (47.1) is **less than** 48, we **cannot** guarantee with 95% confidence that the alloy meets the standard. It implies the mean *could* be as low as 47.1.

- (a)  $\bar{x} = 50, s = 4.06$   
 (b)  $[47.1, 52.9]$  MPa  
 (c) No strong evidence (Lower bound  $< 48$ ).

### 9.5 Difference of Means

**Problem:** Catalyst A vs B. Equal variances assumed.

**Solution:**

(a) **Stats:**

A: {85, 88, 84, 86, 90, 83, 87, 85}.  $n_A = 8, \bar{x}_A = 86.0$ .  $s_A^2 = \frac{1}{7} \sum (x - 86)^2 = \frac{32}{7} \approx 4.57$ .

B: {81, 78, 83, 82, 80, 79, 84, 81}.  $n_B = 8, \bar{x}_B = 81.0$ .  $s_B^2 = \frac{1}{7} \sum (x - 81)^2 = \frac{28}{7} = 4.00$ .

(b) **Pooled Variance:**

$$s_p^2 = \frac{(n_A - 1)s_A^2 + (n_B - 1)s_B^2}{n_A + n_B - 2} = \frac{7(4.57) + 7(4.00)}{14} = \frac{32 + 28}{14} = \frac{60}{14} \approx 4.286$$

$$s_p = \sqrt{4.286} \approx 2.07.$$

(c) **95% CI:**  $df = 14$ .  $t_{0.025,14} = 2.145$ .

$$\begin{aligned}(\bar{x}_A - \bar{x}_B) \pm t \cdot s_p \sqrt{\frac{1}{n_A} + \frac{1}{n_B}} &= (86 - 81) \pm 2.145(2.07) \sqrt{\frac{1}{8} + \frac{1}{8}} \\&= 5 \pm 2.145(2.07)(0.5) \\&= 5 \pm 2.22.\end{aligned}$$

$$CI = [2.78, 7.22].$$

(d) **Conclusion:** The interval lies entirely above 0 (Lower bound  $2.78 > 0$ ). Yes, Catalyst A produces a significantly higher yield.

- (b)  $s_p^2 \approx 4.29$   
 (c)  $[2.78, 7.22]$   
 (d) Yes, significantly higher.

## 9.6 Variance Estimation

**Problem:** Data deviations:  $\{-2, 1, 0, -1, 2, -3, 0, 1, -2, 1, 3, -1\}$ .  $n = 12$ .

**Solution:**

(a) **Sample Variance:**  $\sum x = -1$ .  $\bar{x} = -1/12 \approx -0.083$ .  $s^2 = \frac{\sum x^2 - (\sum x)^2/n}{n-1} = \frac{35 - (-1)^2/12}{11} = \frac{34.917}{11} \approx 3.174$ .

(b) **90% CI for  $\sigma^2$ :**  $\chi^2$  distribution with  $df = 11$ .  $\alpha = 0.10$ .  $\chi_{0.05,11}^2 = 19.675$  (Upper),  $\chi_{0.95,11}^2 = 4.575$  (Lower).

$$\begin{aligned}\left[ \frac{(n-1)s^2}{\chi_{\text{upper}}^2}, \frac{(n-1)s^2}{\chi_{\text{lower}}^2} \right] &= \left[ \frac{11(3.174)}{19.675}, \frac{11(3.174)}{4.575} \right] \\&= \left[ \frac{34.914}{19.675}, \frac{34.914}{4.575} \right] \\&= [1.77, 7.63].\end{aligned}$$

(c) **Within Spec ( $\sigma^2 \leq 4$ )?** The interval extends up to 7.63. Since values  $> 4$  are plausible (inside the CI), we **cannot** confirm with 90% confidence that the machine is within spec.

- (a)  $s^2 \approx 3.17$   
 (b)  $[1.77, 7.63]$   
 (c) No, upper bound  $7.63 > 4$ .

## 9.7 Proportions (A/B Testing)

**Problem:** Red (50/1000) vs Blue (70/1000).

**Solution:**

(a) **Proportions:**  $\hat{p}_R = 0.050$ ,  $\hat{p}_B = 0.070$ .

(b) **99% CI for Diff:**

$$z_{0.005} = 2.576.$$

$$SE = \sqrt{\frac{p_R q_R}{n} + \frac{p_B q_B}{n}} = \sqrt{\frac{0.05(0.95)}{1000} + \frac{0.07(0.93)}{1000}} \approx 0.0106.$$

$$\text{Diff} = 0.070 - 0.050 = 0.020.$$

$$CI = 0.020 \pm 2.576(0.0106) = 0.020 \pm 0.027$$

$$[-0.007, 0.047]$$

(c) **Conclusion:** The interval **contains 0**. This means the difference is not statistically significant at 99% confidence. We cannot definitively say Blue is better.

(b)  $[-0.007, 0.047]$

(c) Contains 0. Difference not significant.

## 9.3 Challenge

### 9.8 Pivotal Quantity Method

**Problem:** Derive T-interval.

**Solution:**

(a) **Pivot:**  $T = \frac{\bar{X} - \mu}{S/\sqrt{n}} \sim t_{n-1}$ . This is pivotal because its distribution ( $t$ -dist) is parameter-free.

(b) **Probability Statement:**

$$P(-t_{\alpha/2} < \frac{\bar{X} - \mu}{S/\sqrt{n}} < t_{\alpha/2}) = 1 - \alpha$$

(c) **Rearrange:** Multiply by  $S/\sqrt{n}$ :

$$-t_{\alpha/2} \frac{S}{\sqrt{n}} < \bar{X} - \mu < t_{\alpha/2} \frac{S}{\sqrt{n}}$$

Subtract  $\bar{X}$ :

$$-\bar{X} - t_{\alpha/2} \frac{S}{\sqrt{n}} < -\mu < -\bar{X} + t_{\alpha/2} \frac{S}{\sqrt{n}}$$

Multiply by -1 (flip inequalities):

$$\bar{X} + t_{\alpha/2} \frac{S}{\sqrt{n}} > \mu > \bar{X} - t_{\alpha/2} \frac{S}{\sqrt{n}}$$

Rewrite:

$$\bar{X} - t_{\alpha/2} \frac{S}{\sqrt{n}} < \mu < \bar{X} + t_{\alpha/2} \frac{S}{\sqrt{n}}$$

Proven.

### 9.9 CI for Gamma Parameter

**Problem:**  $Y = \frac{2}{\beta} \sum X_i \sim \chi^2_{2n\alpha}$ .

**Solution:**

(a) **Validity:**  $Y$  is a function of data ( $\sum X_i$ ) and parameter ( $\beta$ ), and its distribution ( $\chi^2$ ) is known and fixed (independent of  $\beta$ ). Thus, it is a pivot.

(b) **Derive:**

$$P\left(\chi^2_{\text{lower}} < \frac{2\sum X_i}{\beta} < \chi^2_{\text{upper}}\right) = 1 - \alpha.$$

Invert the inequalities:

$$\frac{1}{\chi_{\text{lower}}^2} > \frac{\beta}{2 \sum X_i} > \frac{1}{\chi_{\text{upper}}^2}.$$

Multiply through by  $2 \sum X_i$ :

$$\frac{2 \sum X_i}{\chi_{\text{upper}}^2} < \beta < \frac{2 \sum X_i}{\chi_{\text{lower}}^2}.$$

$$\text{where } \chi_{\text{upper}}^2 = \chi_{\alpha/2}^2 \quad \text{and} \quad \chi_{\text{lower}}^2 = \chi_{1-\alpha/2}^2.$$

- (c) **Application:** Exponential  $\alpha = 1, n = 10 \implies df = 2(10)(1) = 20. \sum x = 500.$   
 $\chi_{0.025,20}^2 = 34.17, \chi_{0.975,20}^2 = 9.59.$

$$CI = \left[ \frac{2(500)}{34.17}, \frac{2(500)}{9.59} \right] = \left[ \frac{1000}{34.17}, \frac{1000}{9.59} \right] = [29.27, 104.28]$$

(b)  $\left[ \frac{2 \sum X_i}{\chi_{\alpha/2}^2}, \frac{2 \sum X_i}{\chi_{1-\alpha/2}^2} \right]$   
(c)  $[29.27, 104.28]$  hours

## 9.10 Prediction Interval vs CI

**Problem:** Error  $X_{n+1} - \bar{X}$ .

**Solution:**

- (a) **Variance:** Since  $X_{n+1}$  is independent of the sample mean  $\bar{X}$ :

$$Var(X_{n+1} - \bar{X}) = Var(X_{n+1}) + Var(\bar{X}) = \sigma^2 + \frac{\sigma^2}{n} = \sigma^2 \left( 1 + \frac{1}{n} \right)$$

- (b) **Formula:** The standard error is  $s\sqrt{1 + 1/n}$ .

$$PI = \bar{X} \pm t_{\alpha/2, n-1} s \sqrt{1 + \frac{1}{n}}$$

(Note: This interval is always wider than the CI because it accounts for the uncertainty of the mean PLUS the variation of the single new data point).

(a)  $\sigma^2(1 + 1/n)$   
(b)  $\bar{X} \pm t \cdot s \sqrt{1 + 1/n}$

## 9.4 Application

### 9.11 Visualizing "Confidence"

**Problem:** Simulation of 100 Intervals.

**Solution:**

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3 from scipy.stats import t
4
5 mu = 50
6 sigma = 10
7 n = 30
8 num_intervals = 100
9 confidence = 0.95
10
11 # Initialize
12 captured_count = 0
13 plt.figure(figsize=(10, 8))
14
15 for i in range(num_intervals):
16     # 1. Generate Sample
17     sample = np.random.normal(mu, sigma, n)
18     x_bar = np.mean(sample)
19     s = np.std(sample, ddof=1)
20
21     # 2. Calculate CI
22     crit_val = t.ppf((1 + confidence) / 2, df=n-1)
23     margin_error = crit_val * (s / np.sqrt(n))
24
25     low = x_bar - margin_error
26     high = x_bar + margin_error
27
28     # 3. Check Capture
29     is_captured = low <= mu <= high
30     if is_captured:
31         captured_count += 1
32         color = 'green'
33     else:
34         color = 'red'
35
36     # 4. Plot
37     plt.plot([low, high], [i, i], color=color, alpha=0.7)
38
39 # Finalize Plot
40 plt.axvline(mu, color='black', linestyle='--', linewidth=2, label='True
    Mean')
41 plt.title(f'100 Confidence Intervals Simulation\nCaptured: {
    captured_count}/100')
42 plt.xlabel('Value')
43 plt.ylabel('Simulation Index')
44 plt.legend()
45 plt.show()
46
47 # Interpretation:
48 # You should see approximately 95 green lines (captured) and 5 red
```



```
lines (missed).  
49 # This demonstrates that "95% Confidence" describes the long-run  
    success rate of the method.
```