# **Confidence Interval Simulation**

# Examining 90% and 95% Coverage

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### **Table of contents**

Overview $\ldots$ $\ldots$ $\ldots$ $\ldots$ $1$
Setup
Simulation Function
Generate Confidence Intervals
Coverage Statistics
Plotting Function
Visualization
90% Confidence Intervals
95% Confidence Intervals
Hit Percentages Plot
Cumulative Hit Percentage Plot
Combined Hit Percentage Plot (90% and 95%) $\dots \dots 9$
Conclusion

#### **Overview**

This simulation demonstrates how confidence intervals behave when repeatedly sampling from a known population. We'll construct both 90% and 95% confidence intervals for men's heights and examine their coverage properties.

### Setup

We'll work with men's heights following a normal distribution with: - Population mean ( ) =  $174.4~\rm cm$  - Population standard deviation ( ) =  $6.57~\rm cm$ 

```
set.seed(123) # For reproducibility
mu <- 174.4 # Population mean
sigma <- 6.57 # Population standard deviation
n <- 8 # Sample size
n_simulations <- 100 # Number of confidence intervals</pre>
```

#### **Simulation Function**

```
simulate_ci <- function(confidence_level) {</pre>
  ci_data <- data.frame(</pre>
    sample_num = 1:n_simulations,
    sample_mean = numeric(n_simulations),
    lower_bound = numeric(n_simulations),
    upper_bound = numeric(n_simulations),
   contains_mu = logical(n_simulations)
  # t-value for given confidence level
  t_value \leftarrow qt((1 + confidence_level)/2, df = n - 1)
  for (i in 1:n_simulations) {
    sample <- rnorm(n, mean = mu, sd = sigma)</pre>
    x_bar <- mean(sample)</pre>
    s <- sd(sample)
    margin <- t_value * s / sqrt(n)</pre>
    ci_data$sample_mean[i] <- x_bar</pre>
    ci_data$lower_bound[i] <- x_bar - margin</pre>
    ci_data$upper_bound[i] <- x_bar + margin</pre>
    ci_data$contains_mu[i] <- (mu >= ci_data$lower_bound[i]) &
                                 (mu <= ci_data$upper_bound[i])</pre>
 }
 return(ci_data)
```

#### **Generate Confidence Intervals**

```
ci_90 <- simulate_ci(0.90)
ci_95 <- simulate_ci(0.95)</pre>
```

### **Coverage Statistics**

```
coverage_90 <- mean(ci_90$contains_mu) * 100
coverage_95 <- mean(ci_95$contains_mu) * 100
cat("90% CI Coverage:", coverage_90, "%\n")

90% CI Coverage: 90 %
cat("95% CI Coverage:", coverage_95, "%\n")

95% CI Coverage: 91 %</pre>
```

### **Plotting Function**

```
plot_ci <- function(ci_data, confidence_level) {
   par(mar = c(5, 4, 4, 2) + 0.1)
   plot(1:n_simulations, ci_data$sample_mean, type = "n",
        ylim = range(c(ci_data$lower_bound, ci_data$upper_bound)),
        xlab = "Sample Number",
        ylab = "Height (cm)",
        main = paste0(confidence_level*100, "% Confidence Intervals"))

# Draw horizontal line at true mean
   abline(h = mu, col = "red", lwd = 2)

# Draw confidence intervals
for (i in 1:n_simulations) {
      col <- ifelse(ci_data$contains_mu[i], "black", "blue")
      segments(i, ci_data$lower_bound[i], i, ci_data$upper_bound[i], col = col)
      points(i, ci_data$sample_mean[i], pch = 16, cex = 0.5, col = col)</pre>
```

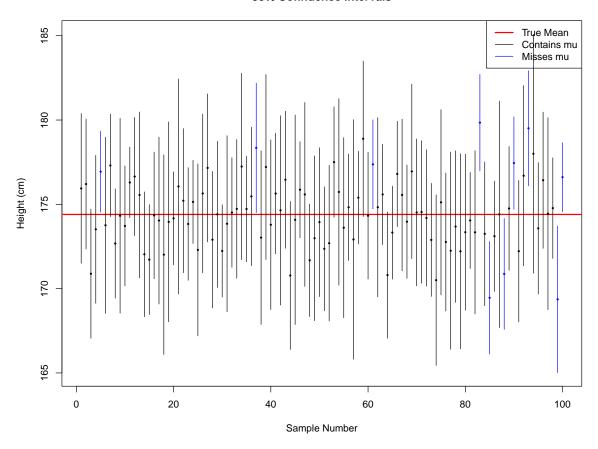
```
legend("topright",
    legend = c("True Mean", "Contains mu", "Misses mu"),
    col = c("red", "black", "blue"),
    lty = c(1, 1, 1),
    lwd = c(2, 1, 1))
}
```

# Visualization

# 90% Confidence Intervals

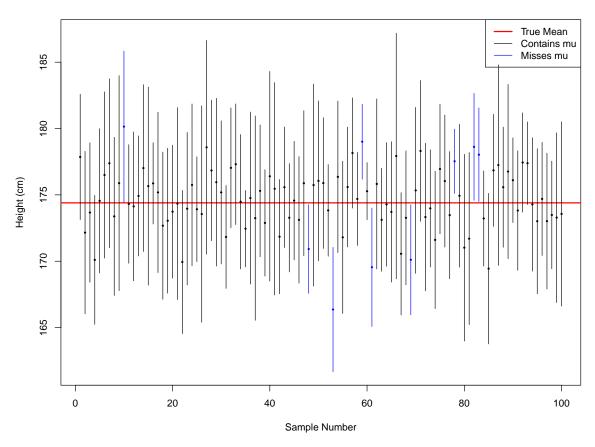
```
par(mfrow = c(1, 1))
plot_ci(ci_90, 0.90)
```

90% Confidence Intervals



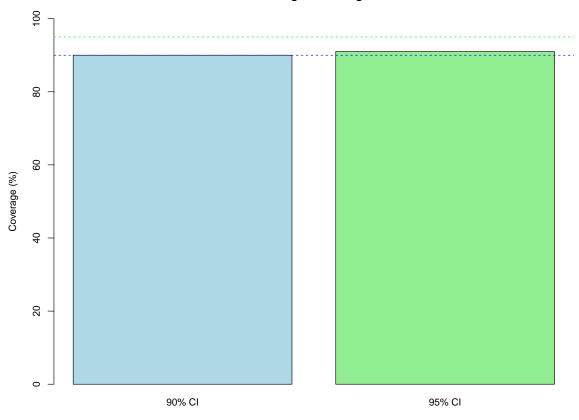
# 95% Confidence Intervals

#### 95% Confidence Intervals



# Hit Percentages Plot

#### **Coverage Percentages**



# **Cumulative Hit Percentage Plot**

```
xaxt = "n")

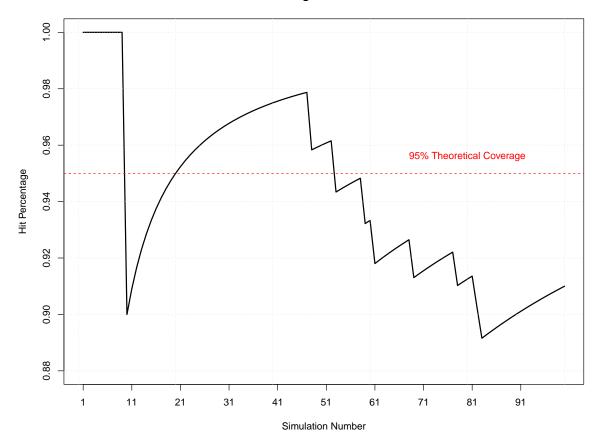
# Add x-axis with specific tick marks
axis(1, at = seq(1, n_simulations, by = 10))

# Add grid
grid(nx = NULL, ny = NULL, col = "lightgray", lty = "dotted")

# Add horizontal line at 0.95
abline(h = 0.95, col = "red", lty = 2)

# Add text label for the theoretical coverage
text(80, 0.953, "95% Theoretical Coverage", col = "red", pos = 3)
```

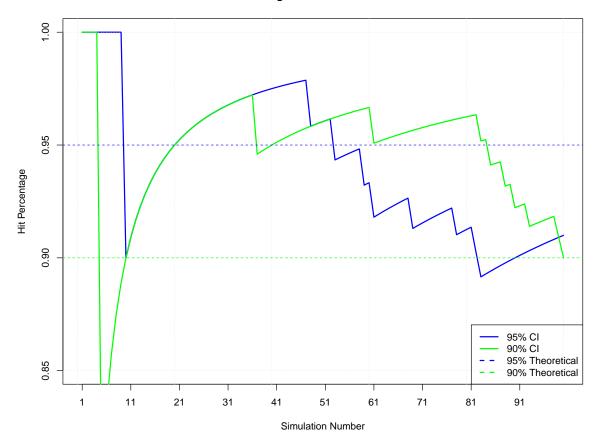
#### **Cumulative Hit Percentage for 95% Confidence Intervals**



### Combined Hit Percentage Plot (90% and 95%)

```
# Calculate cumulative hit percentages for both confidence levels
cumulative_hits_90 <- cumsum(ci_90$contains_mu) / (1:n_simulations)</pre>
cumulative_hits_95 <- cumsum(ci_95$contains_mu) / (1:n_simulations)</pre>
# Create the plot
plot(1:n_simulations, cumulative_hits_95,
     type = "1",
     lwd = 2,
     col = "blue",
     xlab = "Simulation Number",
     ylab = "Hit Percentage",
     main = "Cumulative Hit Percentage for 90% and 95% Confidence Intervals",
     ylim = c(0.85, 1.0),
     xaxt = "n")
# Add x-axis with specific tick marks
axis(1, at = seq(1, n_simulations, by = 10))
# Add the 90% line
lines(1:n_simulations, cumulative_hits_90, col = "green", lwd = 2)
# Add grid
grid(nx = NULL, ny = NULL, col = "lightgray", lty = "dotted")
# Add horizontal lines at theoretical coverage levels
abline(h = 0.95, col = "blue", lty = 2)
abline(h = 0.90, col = "green", lty = 2)
# Add legend
legend("bottomright",
       legend = c("95% CI", "90% CI", "95% Theoretical", "90% Theoretical"),
       col = c("blue", "green", "blue", "green"),
       lty = c(1, 1, 2, 2),
       lwd = 2)
```

#### Cumulative Hit Percentage for 90% and 95% Confidence Intervals



### **Conclusion**

The simulation demonstrates that:

- 1. The 90% confidence intervals capture the true mean approximately 90% of the time
- 2. The 95% confidence intervals capture the true mean approximately 95% of the time
- 3. 95% confidence intervals are wider than 90% confidence intervals, reflecting the higher confidence level
- 4. The coverage rates match our theoretical expectations, validating the confidence interval methodology
- 5. The cumulative hit percentage converges toward the theoretical coverage level as the number of simulations increases