ST502Project

2024-09-29

libraries

```
library(ggplot2)
```

Global variables

```
# Set parameters and define variables for global use
set.seed(42)
N <- 1000  # samples
p_vals <- seq(0.01, 0.99, length.out = 15)  # Range of true probabilities
alpha <- 0.05  # Significance level
B <- 100  # Number of bootstrap samples
z <- qnorm(1 - alpha / 2)  # z value</pre>
```

Wald

```
waldCI <- function(y, n, alpha = 0.05) {</pre>
  if (y == 0) return(c(0, 0))
  if (y == n) return(c(1, 1))
  p_hat <- y / n</pre>
  error <- z * sqrt((p_hat * (1 - p_hat)) / n)
  return(c(p_hat - error, p_hat + error))
}
# Function to check if the true probability is within the confidence interval
check_coverage <- function(ci, p) {</pre>
  return(p >= ci[1] & p <= ci[2])</pre>
}
# Main function for simulation and coverage calculation
simulate_confidence_intervals <- function(N, n, alpha) {</pre>
  coverage_results <- list()</pre>
  for (p in p_vals) {
    y_samples <- rbinom(N, size = n, prob = p)</pre>
    ci_wald <- t(sapply(y_samples, waldCI, n = n, alpha = alpha))</pre>
    wald_coverage <- mean(apply(ci_wald, 1, check_coverage, p = p))</pre>
    coverage_results[[as.character(p)]] <- data.frame(</pre>
      p = p,
      coverage = wald_coverage,
      method = "Wald"
    )
  }
  # Combine all results into a single data frame
```

```
coverage_results_df <- do.call(rbind, coverage_results)</pre>
 return(coverage_results_df)
}
# Sample sizes to iterate over
n_{values} < c(15, 30, 100)
# Loop through each sample size and create separate plots
for (n in n values) {
  # Run simulation with desired parameters
 wald_coverage_results <- simulate_confidence_intervals(N, n)</pre>
  # Plot the coverage probabilities for the Wald interval
 library(ggplot2)
p_plot <-
  ggplot(wald_coverage_results, aes(x = p, y = coverage)) +
   geom_line(color = "blue", size = 1.2) +
   labs(title = paste("Wald Interval Coverage Probability for n =", n),
         x = "True Probability (p)",
         y = "Coverage Probability") +
   ylim(0.7, 1.0) +
   geom_hline(yintercept = 0.95, linetype = "dashed", color = "red") + # 95% target line
   theme minimal() +
   theme(legend.position = "none",
   panel.background = element_rect(fill = "white", color = "white"), # Set panel background to white
          plot.background = element_rect(fill = "white", color = "white"))
  # Save the plot as a PNG file
  ggsave(filename = paste("wald_coverage_n_", n, ".png", sep = ""), plot = p_plot, width = 8, height =
## Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use `linewidth` instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
## Warning: Removed 3 rows containing missing values (`geom_line()`).
## Warning: Removed 2 rows containing missing values (`geom_line()`).
## Removed 2 rows containing missing values (`geom_line()`).
Adjusted Wald
AwaldCI <- function(y, n, alpha) {
 p \leftarrow (y + 2) / (n + 4)
 ci \leftarrow c(p - z * sqrt(p * (1 - p) / (n + 4)),
          p + z * sqrt(p * (1 - p) / (n + 4)))
 return(ci)
# Function to check if the true probability is within the confidence interval
check_coverage <- function(ci, p) {</pre>
 return(p >= ci[1] & p <= ci[2])
```

```
# Main function for simulation and coverage calculation
simulate_Awald_intervals <- function(N, n, alpha) {</pre>
  coverage_results <- list()</pre>
 for (p in p_vals) {
    y_samples <- rbinom(N, size = n, prob = p)</pre>
    ci_Awald <- t(sapply(y_samples, AwaldCI, n = n, alpha = alpha))</pre>
    Awald_coverage <- mean(apply(ci_Awald, 1, check_coverage, p = p))
    coverage_results[[as.character(p)]] <- data.frame(</pre>
     coverage = Awald_coverage,
      method = "Adjusted Wald"
    )
  }
  # Combine all results into a single data frame
  coverage_results_df <- do.call(rbind, coverage_results)</pre>
 return(coverage_results_df)
}
# Set seed for reproducibility
set.seed(123)
# Sample sizes to iterate over
n_{values} < c(15, 30, 100)
# Loop through each sample size and create separate plots
for (n in n_values) {
  # Run simulation with desired parameters
  Awald_coverage_results <- simulate_Awald_intervals(N = 1000, n = n)
  # Plot the coverage probabilities for the Adjusted Wald interval
 library(ggplot2)
p_plot <- ggplot(Awald_coverage_results, aes(x = p, y = coverage)) +</pre>
    geom_line(color = "blue", size = 1.2) +
    labs(title = paste("Adjusted Wald Interval Coverage Probability for n =", n),
         x = "True Probability (p)",
         y = "Coverage Probability") +
    ylim(0.7, 1.0) +
    geom_hline(yintercept = 0.95, linetype = "dashed", color = "red") + # 95% target line
    theme_minimal() +
    theme(legend.position = "none",,
    panel.background = element_rect(fill = "white", color = "white"), # Set panel background to white
          plot.background = element_rect(fill = "white", color = "white"))
  # Save the plot as a PNG file
  ggsave(filename = paste("Awald_coverage_n_", n, ".png", sep = ""), plot = p_plot, width = 8, height =
```

Score

```
scoreCI <- function(y, n, alpha) {</pre>
 p <- y / n
  ci <- c(
    (p + z^2 / (2 * n) - z * sqrt((p * (1 - p) + z^2 / (4 * n)) / n)) / (1 + z^2 / n),
    (p + z^2 / (2 * n) + z * sqrt((p * (1 - p) + z^2 / (4 * n)) / n)) / (1 + z^2 / n)
 return(ci)
}
# Function to check if the true probability is within the confidence interval
check_coverage <- function(ci, p) {</pre>
  return(p >= ci[1] & p <= ci[2])
# Main function for simulation and coverage calculation
simulate_score_intervals <- function(N, n, alpha ) {</pre>
  coverage_results <- list()</pre>
 for (p in p_vals) {
    y_samples <- rbinom(N, size = n, prob = p)</pre>
    ci_score <- t(sapply(y_samples, scoreCI, n, alpha))</pre>
    score_coverage <- mean(apply(ci_score, 1, check_coverage, p = p))</pre>
    coverage results[[as.character(p)]] <- data.frame(</pre>
      coverage = score_coverage,
      method = "Score"
  }
  # Combine all results into a single data frame
  coverage_results_df <- do.call(rbind, coverage_results)</pre>
 return(coverage_results_df)
}
# Sample sizes to iterate over
n_{values} \leftarrow c(15, 30, 100)
# Loop through each sample size and create separate plots
for (n in n_values) {
  # Run simulation with desired parameters
  score_coverage_results <- simulate_score_intervals(N , n )</pre>
  # Plot the coverage probabilities for the Score interval
  library(ggplot2)
  p_plot <- ggplot(score_coverage_results, aes(x = p, y = coverage)) +</pre>
    geom_line(color = "green", size = 1.2) +
    labs(title = paste("Score Interval Coverage Probability for n =", n),
         x = "True Probability (p)",
         y = "Coverage Probability") +
```

```
ylim(0.7, 1.0) +
    geom_hline(yintercept = 0.95, linetype = "dashed", color = "red") + # 95% target line
    theme_minimal() +
    theme(legend.position = "none",,
    panel.background = element_rect(fill = "white", color = "white"), # Set panel background to white
          plot.background = element_rect(fill = "white", color = "white"))
  # Save the plot as a PNG file
  ggsave(filename = paste("score_coverage_n_", n, ".png", sep = ""), plot = p_plot, width = 8, height =
Clopper Pearson (Exact)
exactCI <- function(y, n, alpha) {</pre>
  if (y == 0) return(c(0, 1 - (alpha / 2)^(1/n)))
  if (y == n) return(c((alpha / 2)^(1/n), 1))
 lower <- 1 / (1 + (n - y + 1) / (y * qf(1 - alpha / 2, 2 * y, 2 * (n - y + 1), lower.tail = FALSE)))
 upper <- 1 / (1 + (n - y) / ((y + 1) * qf(alpha / 2, 2 * (y + 1), 2 * (n - y), lower.tail = FALSE)))
 return(c(lower, upper))
}
# Function to check if the true probability is within the confidence interval
check_coverage <- function(ci, p) {</pre>
 return(p >= ci[1] & p <= ci[2])
}
# Main function for simulation and coverage calculation
simulate exact intervals <- function(N, n, alpha) {</pre>
 p_{vals} \leftarrow seq(0.01, 0.99, length.out = 100)
  coverage results <- list()</pre>
 for (p in p_vals) {
    y_samples <- rbinom(N, size = n, prob = p)</pre>
    ci_exact <- t(sapply(y_samples, exactCI, n, alpha))</pre>
    exact_coverage <- mean(apply(ci_exact, 1, check_coverage, p = p))</pre>
    coverage_results[[as.character(p)]] <- data.frame(</pre>
      p = p,
      coverage = exact_coverage,
      method = "Exact"
  }
  # Combine all results into a single data frame
  coverage_results_df <- do.call(rbind, coverage_results)</pre>
 return(coverage_results_df)
}
# Sample sizes to iterate over
n values \leftarrow c(15, 30, 100)
```

```
# Loop through each sample size and create separate plots
for (n in n_values) {
  # Run simulation with desired parameters
  exact_coverage_results <- simulate_exact_intervals(N, n, alpha)</pre>
  # Plot the coverage probabilities for the Exact (Clopper-Pearson) interval
 library(ggplot2)
 p_plot <- ggplot(exact_coverage_results, aes(x = p, y = coverage)) +</pre>
    geom_line(color = "purple", size = 1.2) +
    labs(title = paste("Exact Interval (Clopper-Pearson) Coverage Probability for n =", n),
         x = "True Probability (p)",
         y = "Coverage Probability") +
    ylim(0.7, 1.0) +
    geom_hline(yintercept = 0.95, linetype = "dashed", color = "red") + # 95% target line
    theme minimal() +
    theme(legend.position = "none",
    panel.background = element_rect(fill = "white", color = "white"), # Set panel background to white
          plot.background = element_rect(fill = "white", color = "white"))
  # Save the plot as a PNG file
  ggsave(filename = paste("exact_coverage_n_", n, ".png", sep = ""), plot = p_plot, width = 8, height =
}
Bootstrap based Intervals for raw
bootstrap_percentileCI <- function(y, n, B , alpha) {</pre>
  # Early exit for edge cases
  if (y == 0) return(list(percentile = c(0, 0)))
  if (y == n) return(list(percentile = c(1, 1)))
  p_hat <- y / n
  # Generate bootstrap samples for the mean
  boot_samples <- replicate(B, mean(rbinom(n, size = 1, prob = p_hat)))</pre>
  # Percentile-based confidence interval
  ci_percentile <- quantile(boot_samples, probs = c(alpha / 2, 1 - alpha / 2), na.rm = TRUE)
  # Return the confidence interval as a named list
 return(list(percentile = ci_percentile))
}
# Function to check coverage
check coverage <- function(ci, p) {</pre>
 return(p >= ci[1] & p <= ci[2])
}
# Main function for Bootstrap Percentile interval
simulate_bootstrap_percentile_intervals <- function(N , n, alpha , B ) {</pre>
 p_{vals} \leftarrow seq(0.01, 0.99, length.out = 15)
  coverage_results <- list()</pre>
 for (p in p_vals) {
    y_samples <- rbinom(N, size = n, prob = p)</pre>
```

```
ci_percentile_list <- vector("list", length(y_samples))</pre>
    for (i in seq_along(y_samples)) {
      ci_percentile_list[[i]] <- bootstrap_percentileCI(y_samples[i], n, B = B, alpha = alpha)$percenti
    ci_percentile_matrix <- do.call(rbind, ci_percentile_list)</pre>
    percentile_coverage <- mean(apply(ci_percentile_matrix, 1, check_coverage, p = p))</pre>
    coverage_results[[as.character(p)]] <- data.frame(</pre>
      p = p,
      coverage = percentile_coverage,
      method = "Bootstrap Percentile"
    )
  }
  # Combine all results into a single data frame
  coverage_results_df <- do.call(rbind, coverage_results)</pre>
 return(coverage_results_df)
}
# Sample sizes to iterate over
n_{values} \leftarrow c(15, 30, 100)
# Loop through each sample size and create separate plots
for (n in n values) {
  # Run simulation with desired parameters
  bootstrap_percentile_results <- simulate_bootstrap_percentile_intervals(N , n, alpha, B )
  # Plot the coverage probabilities for the Bootstrap Percentile interval
  library(ggplot2)
  p_plot <- ggplot(bootstrap_percentile_results, aes(x = p, y = coverage)) +</pre>
    geom_line(color = "orange", size = 1.2) +
    labs(title = paste("Bootstrap Percentile Interval Coverage Probability for n =", n),
         x = "True Probability (p)",
         y = "Coverage Probability") +
    ylim(0.7, 1.0) +
    geom_hline(yintercept = 0.95, linetype = "dashed", color = "red") + # 95% target line
    theme_minimal() +
    theme(legend.position = "none",
    panel.background = element_rect(fill = "white", color = "white"), # Set panel background to white
          plot.background = element_rect(fill = "white", color = "white"))
  # Save the plot as a PNG file
  ggsave(filename = paste("bootstrap_percentile_coverage_n_", n, ".png", sep = ""), plot = p_plot, widt
## Warning: Removed 3 rows containing missing values (`geom_line()`).
## Warning: Removed 2 rows containing missing values (`geom_line()`).
## Removed 2 rows containing missing values (`geom_line()`).
```

Bootstrap t Interval

```
bootstrap_tCI <- function(y, n, B , alpha ) {</pre>
  if (y == 0) return(c(0, 0))
  if (y == n) return(c(1, 1))
 p_hat <- y / n</pre>
  # Generate bootstrap samples
  boot_samples <- replicate(B, mean(rbinom(n, size = 1, prob = p_hat)))</pre>
  # Compute t-statistics
 t_stat <- (boot_samples - p_hat) / sqrt((p_hat * (1 - p_hat)) / n)
  # T-interval confidence interval
 ci_t <- p_hat + quantile(t_stat, probs = c(alpha / 2, 1 - alpha / 2), na.rm = TRUE) * sqrt((p_hat * (</pre>
 return(ci_t)
# Parameters
bootstrap_t_results <- data.frame()</pre>
# Sample sizes to iterate over
n_{values} < c(15, 30, 100)
# Bootstrap T-Interval Coverage Simulation
for (n in n_values) {
 for (p in p_vals) {
  y_samples <- rbinom(B, size = n, prob = p)</pre>
 ci_t_interval <- t(sapply(y_samples, bootstrap_tCI, n = n, B, alpha = alpha))</pre>
 t_interval_coverage <- mean(apply(ci_t_interval, 1, check_coverage, p = p))</pre>
  # Store results for bootstrap t-interval
  bootstrap_t_results <- rbind(bootstrap_t_results, data.frame(</pre>
    p = p,
    coverage = t_interval_coverage,
    method = "bootstrap_t_interval"
    ))
 }
}
# Plot for Bootstrap T-Interval
for(n in n_values) {
p_plot <- ggplot(bootstrap_t_results, aes(x = p, y = coverage, color = method, linetype = method)) +</pre>
  geom_line(size = 1.2) +
  labs(title = paste("Bootstrap T-Interval Coverage (n = ", n, ")"),
       x = "True Probability (p)",
       y = "Coverage Probability") +
  geom_hline(yintercept = 0.95, linetype = "dashed", color = "red") + # 95% target line
  theme_minimal() + ylim(0.7, 1.0) +
  theme(legend.position = "bottom",
    panel.background = element_rect(fill = "white", color = "white"), # Set panel background to white
          plot.background = element_rect(fill = "white", color = "white"))
```

```
bootstrap_t_results
ci t interval
# Save the plot as a PNG file
  ggsave(filename = paste("bootstrap_t_coverage_n_", n, ".png", sep = ""), plot = p_plot, width = 8, he
## Warning: Removed 6 rows containing missing values (`geom_line()`).
## Removed 6 rows containing missing values (`geom_line()`).
## Removed 6 rows containing missing values (`geom_line()`).
Proportion that miss above or below for bootstrapped
# Load necessary library
library(ggplot2)
# Function to check coverage details
check_coverage_details <- function(ci, p) {</pre>
  if (p \ge ci[1] \& p \le ci[2]) {
   return(c(capture = 1, miss_above = 0, miss_below = 0)) # Captures the true value
 } else if (p < ci[1]) {</pre>
    return(c(capture = 0, miss_above = 1, miss_below = 0)) # Misses above
 } else {
    return(c(capture = 0, miss_above = 0, miss_below = 1)) # Misses below
}
# Initialize a data frame to store results
bootstrap_coverage_results <- data.frame()</pre>
# Function for Bootstrap t Interval
bootstrap_tCI <- function(y, n, B = 100, alpha = 0.05) {</pre>
 if (y == 0) return(c(0, 0))
 if (y == n) return(c(1, 1))
 p_hat <- y / n
  # Generate bootstrap samples
  boot_samples <- replicate(B, mean(rbinom(n, size = 1, prob = p_hat)))</pre>
  # Compute t-statistics
 t_stat <- (boot_samples - p_hat) / sqrt((p_hat * (1 - p_hat)) / n)
  # T-interval confidence interval
 ci_t <- p_hat + quantile(t_stat, probs = c(alpha / 2, 1 - alpha / 2), na.rm = TRUE) * sqrt((p_hat * (</pre>
 return(ci_t)
}
# Generate samples and compute coverage probabilities for both bootstrap methods
for (p in p_vals) {
 y_samples <- rbinom(100, size = n, prob = p) # Generate random samples</pre>
  # Calculate bootstrap CIs for percentile
```

```
ci_bootstrap_percentile <- lapply(y_samples, bootstrap_percentileCI, n = n, B = B, alpha = alpha)
  ci_percentile <- do.call(rbind, lapply(ci_bootstrap_percentile, function(x) x$percentile))</pre>
  # Calculate bootstrap CIs for t interval
  ci_bootstrap_t <- t(sapply(y_samples, bootstrap_tCI, n = n, B = B, alpha = alpha))</pre>
  # Coverage and miss probabilities for percentile intervals
  percentile_coverage_details <- t(apply(ci_percentile, 1, check_coverage_details, p = p))</pre>
  percentile_coverage <- colMeans(percentile_coverage_details)</pre>
  # Coverage and miss probabilities for t intervals
  t_coverage_details <- t(apply(ci_bootstrap_t, 1, check_coverage_details, p = p))</pre>
  t_coverage <- colMeans(t_coverage_details)</pre>
  # Store the results for both methods
  bootstrap_coverage_results <- rbind(bootstrap_coverage_results,</pre>
                                       data.frame(
                                         p = p,
                                         coverage = percentile_coverage[1], # Capture rate
                                         miss_above = percentile_coverage[2], # Miss above
                                         miss_below = percentile_coverage[3], # Miss below
                                         method = "bootstrap_percentile"
                                       ),
                                       data.frame(
                                         p = p,
                                         coverage = t_coverage[1], # Capture rate
                                         miss_above = t_coverage[2], # Miss above
                                         miss_below = t_coverage[3], # Miss below
                                         method = "bootstrap_t_interval"
                                       ))
}
# Display the results
bootstrap_coverage_results
```

```
##
               p coverage miss_above miss_below
                                                              method
## capture
            0.01
                     0.59
                                0.00
                                           0.41 bootstrap_percentile
## capture1 0.01
                     0.58
                                0.01
                                           0.41 bootstrap_t_interval
                     0.93
## capture2 0.08
                                0.02
                                           0.05 bootstrap_percentile
## capture3 0.08
                     0.94
                                0.02
                                           0.04 bootstrap_t_interval
## capture4 0.15
                     0.89
                                0.05
                                           0.06 bootstrap_percentile
                     0.91
## capture5 0.15
                                0.04
                                           0.05 bootstrap_t_interval
                                           0.02 bootstrap_percentile
## capture6 0.22
                     0.96
                                0.02
## capture7 0.22
                     0.94
                                0.03
                                           0.03 bootstrap t interval
## capture8 0.29
                     0.93
                                0.04
                                           0.03 bootstrap_percentile
## capture9 0.29
                     0.91
                                0.05
                                           0.04 bootstrap_t_interval
                     0.93
                                0.03
## capture10 0.36
                                           0.04 bootstrap_percentile
## capture11 0.36
                     0.91
                                0.04
                                           0.05 bootstrap_t_interval
## capture12 0.43
                     0.98
                                0.00
                                           0.02 bootstrap_percentile
## capture13 0.43
                     0.97
                                0.01
                                           0.02 bootstrap_t_interval
                     0.90
                                0.04
                                           0.06 bootstrap_percentile
## capture14 0.50
## capture15 0.50
                     0.94
                                0.03
                                           0.03 bootstrap_t_interval
                     0.91
                                0.06
## capture16 0.57
                                           0.03 bootstrap_percentile
## capture17 0.57
                     0.93
                                0.04
                                           0.03 bootstrap_t_interval
```

```
## capture18 0.64
                      0.96
                                 0.01
                                             0.03 bootstrap_percentile
## capture19 0.64
                      0.95
                                 0.02
                                             0.03 bootstrap_t_interval
                                             0.02 bootstrap percentile
## capture20 0.71
                      0.93
                                 0.05
## capture21 0.71
                      0.95
                                 0.03
                                            0.02 bootstrap_t_interval
## capture22 0.78
                      0.96
                                 0.03
                                            0.01 bootstrap_percentile
## capture23 0.78
                      0.97
                                 0.03
                                            0.00 bootstrap t interval
## capture24 0.85
                                 0.06
                                            0.01 bootstrap percentile
                      0.93
## capture25 0.85
                      0.93
                                 0.05
                                            0.02 bootstrap_t_interval
## capture26 0.92
                      0.89
                                 0.08
                                            0.03 bootstrap_percentile
## capture27 0.92
                      0.91
                                 0.07
                                            0.02 bootstrap_t_interval
## capture28 0.99
                      0.64
                                 0.36
                                             0.00 bootstrap_percentile
## capture29 0.99
                      0.63
                                 0.36
                                             0.01 bootstrap_t_interval
```

Proportion that miss above and below for Wald

```
# Initialize a data frame to store the results for Wald intervals
wald_coverage_results <- data.frame()</pre>
# Loop over the true probability values to compute coverage and miss rates for the Wald interval
for (p in p_vals) {
  y_samples <- rbinom(1000, size = n, prob = p) # Generate random samples
  ci_wald <- t(sapply(y_samples, waldCI, n = n, alpha = alpha)) # Wald CI for each sample
  # Check coverage and proportion of misses for Wald intervals
  wald_coverage_details <- t(apply(ci_wald, 1, check_coverage_details, p = p))</pre>
  wald_coverage <- colMeans(wald_coverage_details)</pre>
  # Store the results for Wald Interval (without average length)
  wald_coverage_results <- rbind(wald_coverage_results, data.frame(</pre>
   p = p,
    coverage = wald_coverage[1], # Capture rate
   miss_above = wald_coverage[2], # Miss above
   miss_below = wald_coverage[3], # Miss below
   method = "Wald"
  ))
}
# View the Wald interval results
wald_coverage_results
```

```
p coverage miss_above miss_below method
                                                   Wald
## capture
             0.01
                     0.620
                                0.001
                                           0.379
## capture1 0.08
                     0.908
                                0.017
                                           0.075
                                                   Wald
## capture2 0.15
                     0.928
                                0.017
                                           0.055
                                                   Wald
                                           0.055
## capture3 0.22
                     0.935
                                0.010
                                                   Wald
## capture4 0.29
                     0.928
                                           0.055
                                                   Wald
                                0.017
## capture5 0.36
                     0.939
                                0.029
                                           0.032
                                                   Wald
## capture6 0.43
                     0.951
                                0.023
                                           0.026
                                                   Wald
## capture7 0.50
                     0.939
                                0.022
                                           0.039
                                                   Wald
## capture8 0.57
                     0.959
                                0.020
                                           0.021
                                                   Wald
                     0.940
                                0.033
                                           0.027
                                                   Wald
## capture9 0.64
## capture10 0.71
                     0.930
                                0.048
                                           0.022
                                                   Wald
## capture11 0.78
                     0.931
                                0.054
                                           0.015
                                                   Wald
## capture12 0.85
                     0.938
                                0.051
                                           0.011
                                                   Wald
## capture13 0.92
                     0.921
                                0.068
                                           0.011
                                                   Wald
```

```
## capture14 0.99
                     0.642
                                0.358
                                           0.000
                                                   Wald
Proportion that miss above and below for Adj Wald
# Initialize a data frame to store the results for Adjusted Wald intervals
Awald_coverage_results <- data.frame()</pre>
# Loop over the true probability values to compute coverage and miss rates for the Adjusted Wald interv
for (p in p_vals) {
  y samples <- rbinom(1000, size = n, prob = p) # Generate random samples
  ci_Awald <- t(sapply(y_samples, AwaldCI, n = n, alpha = alpha)) # Adjusted Wald CI for each sample
  # Check coverage and proportion of misses for Adjusted Wald intervals
  Awald_coverage_details <- t(apply(ci_Awald, 1, check_coverage_details, p = p))
  Awald_coverage <- colMeans(Awald_coverage_details)</pre>
  # Store the results for Adjusted Wald Interval (without average length)
  Awald_coverage_results <- rbind(Awald_coverage_results, data.frame(</pre>
   p = p,
   coverage = Awald_coverage[1], # Capture rate
   miss_above = Awald_coverage[2], # Miss above
   miss_below = Awald_coverage[3], # Miss below
   method = "AWald"
  ))
}
# View the Adjusted Wald interval results
Awald_coverage_results
                p coverage miss_above miss_below method
## capture
             0.01
                     0.979
                                0.021
                                           0.000 AWald
                     0.965
## capture1 0.08
                                0.021
                                           0.014 AWald
## capture2 0.15
                     0.959
                                0.029
                                           0.012 AWald
                                0.027
                                           0.018 AWald
## capture3 0.22
                     0.955
                                           0.015 AWald
## capture4 0.29
                     0.956
                                0.029
                                0.030
                                           0.033 AWald
## capture5 0.36
                     0.937
## capture6 0.43
                     0.945
                                0.027
                                           0.028 AWald
                                           0.024 AWald
## capture7 0.50
                     0.945
                                0.031
## capture8 0.57
                     0.936
                                0.039
                                           0.025 AWald
                                0.024
                                           0.029 AWald
## capture9 0.64
                     0.947
                                           0.025 AWald
## capture10 0.71
                     0.962
                                0.013
                                           0.023 AWald
## capture11 0.78
                     0.966
                                0.011
## capture12 0.85
                     0.963
                                0.012
                                           0.025 AWald
## capture13 0.92
                     0.948
                                0.014
                                           0.038 AWald
## capture14 0.99
                                0.000
                                           0.010 AWald
                     0.990
Proportion that miss above and below for Score
# Initialize a data frame to store the results for Score intervals
score_coverage_results <- data.frame()</pre>
# Loop over the true probability values to compute coverage and miss rates for the Score interval
for (p in p_vals) {
  y_samples <- rbinom(1000, size = n, prob = p) # Generate random samples</pre>
  ci_score <- t(sapply(y_samples, scoreCI, n = n, alpha = alpha)) # Score CI for each sample
```

```
# Check coverage and proportion of misses for Score intervals
  score_coverage_details <- t(apply(ci_score, 1, check_coverage_details, p = p))</pre>
  score_coverage <- colMeans(score_coverage_details)</pre>
  # Store the results for Score Interval (without average length)
  score_coverage_results <- rbind(score_coverage_results, data.frame(</pre>
   p = p,
   coverage = score_coverage[1], # Capture rate
   miss_above = score_coverage[2], # Miss above
   miss_below = score_coverage[3], # Miss below
   method = "Score"
 ))
}
# View the Score interval results
score_coverage_results
##
                p coverage miss_above miss_below method
## capture
             0.01
                     0.923
                                0.077
                                           0.000 Score
## capture1 0.08
                     0.957
                                0.034
                                           0.009 Score
## capture2 0.15
                     0.933
                                0.036
                                           0.031 Score
                                           0.016 Score
                     0.965
                                0.019
```

capture3 0.22 ## capture4 0.29 0.953 0.024 0.023 Score ## capture5 0.36 0.941 0.036 0.023 Score ## capture6 0.43 0.939 0.036 Score 0.025 ## capture7 0.50 0.023 Score 0.947 0.030 ## capture8 0.57 0.028 0.022 Score 0.950 ## capture9 0.64 0.963 0.019 0.018 Score ## capture10 0.71 0.940 0.030 0.030 Score ## capture11 0.78 0.960 0.018 0.022 Score ## capture12 0.85 0.932 0.038 Score 0.030 0.010 0.043 Score ## capture13 0.92 0.947 0.079 Score ## capture14 0.99 0.921 0.000

Proportion that miss above and below for Clopper Pearson

```
# Initialize a data frame to store the results for Exact intervals
exact_coverage_results <- data.frame()

# Loop over the true probability values to compute coverage and miss rates for the Exact interval
for (p in p_vals) {
    y_samples <- rbinom(1000, size = n, prob = p) # Generate random samples
    ci_exact <- t(sapply(y_samples, exactCI, n = n, alpha = alpha)) # Exact CI for each sample

# Check coverage and proportion of misses for Exact intervals
    exact_coverage_details <- t(apply(ci_exact, 1, check_coverage_details, p = p))
    exact_coverage <- colMeans(exact_coverage_details)

# Store the results for Clopper-Pearson (Exact) Interval (without average length)
exact_coverage_results <- rbind(exact_coverage_results, data.frame(
    p = p,
        coverage = exact_coverage[1], # Capture rate
    miss_above = exact_coverage[2], # Miss above
    miss_below = exact_coverage[3], # Miss below
    method = "Exact"</pre>
```

```
))
}
# View the Exact interval results
exact_coverage_results
##
                p coverage miss_above miss_below method
## capture
             0.01
                      0.987
                                 0.013
                                            0.000 Exact
## capture1 0.08
                      0.976
                                 0.016
                                            0.008 Exact
## capture2 0.15
                      0.975
                                 0.017
                                            0.008 Exact
## capture3 0.22
                      0.973
                                 0.017
                                            0.010 Exact
## capture4 0.29
                      0.968
                                 0.019
                                            0.013 Exact
## capture5 0.36
                                 0.013
                                            0.035 Exact
                      0.952
## capture6 0.43
                      0.963
                                 0.026
                                            0.011 Exact
## capture7 0.50
                      0.956
                                 0.021
                                            0.023 Exact
## capture8 0.57
                      0.960
                                 0.015
                                            0.025 Exact
## capture9 0.64
                      0.967
                                 0.019
                                            0.014 Exact
                                            0.028 Exact
## capture10 0.71
                      0.956
                                 0.016
## capture11 0.78
                      0.963
                                 0.011
                                            0.026 Exact
## capture12 0.85
                      0.968
                                 0.012
                                            0.020 Exact
                                            0.011 Exact
## capture13 0.92
                      0.976
                                 0.013
## capture14 0.99
                      0.994
                                 0.000
                                            0.006 Exact
Average widths
library(ggplot2)
# Function to calculate average length of confidence intervals
calculate_average_length <- function(N, n, alpha) {</pre>
  lengths <- data.frame(p = numeric(), length = numeric(), method = character())</pre>
  for (p in p_vals) {
    y_samples <- rbinom(N, size = n, prob = p)</pre>
    # Calculate lengths for each method
    ci_wald <- t(sapply(y_samples, waldCI, n = n, alpha = alpha))</pre>
    lengths <- rbind(lengths, data.frame(p = p, length = ci_wald[, 2] - ci_wald[, 1], method = "Wald"))</pre>
    ci_Awald <- t(sapply(y_samples, AwaldCI, n = n, alpha = alpha))</pre>
    lengths <- rbind(lengths, data.frame(p = p, length = ci_Awald[, 2] - ci_Awald[, 1], method = "Adjus"</pre>
    ci_score <- t(sapply(y_samples, scoreCI, n = n, alpha = alpha))</pre>
    lengths <- rbind(lengths, data.frame(p = p, length = ci_score[, 2] - ci_score[, 1], method = "Score</pre>
    ci_exact <- t(sapply(y_samples, exactCI, n = n, alpha = alpha))</pre>
    lengths <- rbind(lengths, data.frame(p = p, length = ci_exact[, 2] - ci_exact[, 1], method = "Exact</pre>
    ci_percentile <- sapply(y_samples, function(y) bootstrap_percentileCI(y, n, B, alpha) percentile)
    lengths <- rbind(lengths, data.frame(p = p, length = ci_percentile[2, ] - ci_percentile[1, ], metho</pre>
    ci_t <- sapply(y_samples, function(y) bootstrap_tCI(y, n, B, alpha))</pre>
    lengths <- rbind(lengths, data.frame(p = p, length = ci_t[2, ] - ci_t[1, ], method = "Bootstrap T")</pre>
 # Calculate average lengths
```

```
average_lengths <- aggregate(length ~ p + method, data = lengths, FUN = mean)
  return(average_lengths)
}
# Sample sizes to iterate over
n_{values} < c(15, 30, 100)
# Loop through each sample size and create plots
for (n in n values) {
  average_lengths_results <- calculate_average_length(N = 1000, n = n, alpha = 0.05)
  # Create the plot
  plot <- ggplot(average_lengths_results, aes(x = p, y = length, color = method)) +</pre>
    geom_smooth(se = FALSE, size = 1.2) +
    labs(title = paste("Average Length of Confidence Intervals for n = ", n),
         x = "True Probability (p)",
         y = "Average Length of CIs") +
    ylim(0.0, 0.5) +
    scale_x_continuous(breaks = seq(0, 1, by = 0.25)) +
    theme_minimal() +
    theme(legend.position = "bottom",
    panel.background = element_rect(fill = "white", color = "white"), # Set panel background to white
          plot.background = element_rect(fill = "white", color = "white"))
  # Save the plot
  ggsave(filename = paste("average_length_CI_n_", n, ".png", sep = ""), plot = plot, width = 8, height
}
## `geom_smooth()` using method = 'loess' and formula = 'y ~ x'
## Warning: Removed 3 rows containing non-finite values (`stat_smooth()`).
## Warning: Removed 18 rows containing missing values (`geom_smooth()`).
## `geom_smooth()` using method = 'loess' and formula = 'y ~ x'
## `geom_smooth()` using method = 'loess' and formula = 'y ~ x'
Calculate standard errors for average widths
# Function to calculate average length and standard error of confidence intervals
calculate_average_length_with_se <- function(N, n, alpha) {</pre>
  lengths <- data.frame(p = numeric(), length = numeric(), method = character())</pre>
  for (p in p_vals) {
    y_samples <- rbinom(N, size = n, prob = p)</pre>
    # Calculate lengths for each method
    ci_wald <- t(sapply(y_samples, waldCI, n = n, alpha = alpha))</pre>
    lengths <- rbind(lengths, data.frame(p = p, length = ci_wald[, 2] - ci_wald[, 1], method = "Wald"))</pre>
    ci_Awald <- t(sapply(y_samples, AwaldCI, n = n, alpha = alpha))</pre>
    lengths <- rbind(lengths, data.frame(p = p, length = ci_Awald[, 2] - ci_Awald[, 1], method = "Adjus"</pre>
    ci_score <- t(sapply(y_samples, scoreCI, n = n, alpha = alpha))</pre>
    lengths <- rbind(lengths, data.frame(p = p, length = ci_score[, 2] - ci_score[, 1], method = "Score</pre>
```

```
ci_exact <- t(sapply(y_samples, exactCI, n = n, alpha = alpha))</pre>
    lengths <- rbind(lengths, data.frame(p = p, length = ci_exact[, 2] - ci_exact[, 1], method = "Exact</pre>
    ci_percentile <- sapply(y_samples, function(y) bootstrap_percentileCI(y, n, B, alpha) percentile)
    lengths <- rbind(lengths, data.frame(p = p, length = ci_percentile[2, ] - ci_percentile[1, ], metho</pre>
    ci_t <- sapply(y_samples, function(y) bootstrap_tCI(y, n, B, alpha))</pre>
    lengths <- rbind(lengths, data.frame(p = p, length = ci t[2, ] - ci t[1, ], method = "Bootstrap T")</pre>
  }
  # Calculate average lengths and standard errors
  average_lengths <- aggregate(length ~ method, data = lengths, FUN = mean)
  standard_errors <- aggregate(length ~ method, data = lengths, FUN = function(x) sd(x) / sqrt(N))
  # Combine average lengths and standard errors
  combined_results <- merge(average_lengths, standard_errors, by = "method")</pre>
  colnames(combined_results) <- c("method", "avg_length", "se_length")</pre>
  return(combined_results)
# Initialize a table to store SE results with the correct structure
se_table <- data.frame(Method = character(6), n_15 = numeric(6), n_30 = numeric(6), n_100 = numeric(6))</pre>
# Sample sizes to iterate over
n_{values} \leftarrow c(15, 30, 100)
# Loop through each sample size and store the SE results
for (n in n_values) {
  results <- calculate_average_length_with_se(N = 1000, n = n, alpha = 0.05)
  # Assign the SE results to the appropriate column in se_table
  if (n == 15) {
    se_table$n_15 <- results$se_length
  } else if (n == 30) {
    se_table$n_30 <- results$se_length
  } else if (n == 100) {
    se_table$n_100 <- results$se_length
  # Store the method names in the first column (this only needs to be done once)
  se_table$Method <- results$method</pre>
# Print the table
print(se_table)
##
                                  n_{15}
                                              n_30
            Adjusted Wald 0.001939769 0.001869174 0.001364342
## 1
## 2 Bootstrap Percentile 0.005400166 0.003499173 0.001740917
              Bootstrap T 0.005404468 0.003489766 0.001740458
## 4
                    Exact 0.003397251 0.002676765 0.001590292
## 5
                    Score 0.002755010 0.002318041 0.001491608
```