

# P8160 Simulation Project - Hierarchical Logistic Model for Multicenter Clinical Trial

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## Introduction

*multicenter clinical trial???*

In this project, our objective is to design a simulation study that estimates the population-level probability of adverse events

We are interested in estimating the overall (population-level) probability of an adverse event, marginalizing over both  $X$  (across all patients) and the random effect distribution  $b$  (across all clinics)

## Statistical Methods

*overall description of what we did, the code in this section specifies functions*

### 1. Initialization and Function Definitions

We specified the clinic random effect ( $b$ ) as  $b \sim t_5$  and the patient-level covariates ( $X$ ) as  $X \sim \text{Gamma}(2, 1)$ .

```
# logistic function
logistic <- function(z) {
  1 / (1 + exp(-z))
}

# Set model parameters
alpha <- 0      # Intercept
beta  <- 1      # Covariate coefficient

# Define random generators for non-normal distributions
#               of random effect b and covariate x
# Example: b ~ t_5 (heavy-tailed distribution)
r_b <- function(n) {
  rt(n, df = 5)
}

# Example: x ~ Gamma(shape=2, rate=1)
r_x <- function(n) {
  rgamma(n, shape = 2, rate = 1)
}
```

### 2. Sampling from Simple Monte Carlo (MC)

```
simpleMC <- function(N) {
  # Sample b from f(b)
```

```

b_samp <- r_b(N)
# Sample x from f(x)
x_samp <- r_x(N)

# Compute p_i = logistic(alpha + b_i + beta*x_i)
p_values <- logistic(alpha + b_samp + beta * x_samp)

# Simple MC estimate is the average of p_values
est <- mean(p_values)

return(list(est = est))
}

```

### 3. Designing Control Variate (CV)

```

# 3.1 Define a control variate g(b, x)
# It should be correlated with the target function but have a known or easily
# approximated expectation. As an example, we use a probit-based function.

g_function <- function(b, x) {
  # Example: replace logistic with pnorm (probit)
  z <- alpha + 0.8 * b + 0.8 * x # 0.8 is a heuristic factor here
  return(pnorm(z))
}

# 3.2 Approximate E[g] using a large-scale MC, since we might not have a closed-form
approx_Eg <- function(N = 5e5) {
  b_samp <- r_b(N)
  x_samp <- r_x(N)
  mean(g_function(b_samp, x_samp))
}

# 3.3 Control variate MC estimate
cvMC <- function(N, Eg_true = NULL) {
  if(is.null(Eg_true)) {
    # If not provided, approximate E[g] with a bigger MC sample
    Eg_true <- approx_Eg(N = 5e5)
  }
  # Sample from original distributions
  b_samp <- r_b(N)
  x_samp <- r_x(N)

  # Target function p(b,x)
  p_vals <- logistic(alpha + b_samp + beta * x_samp)
  # Control variate g(b,x)
  g_vals <- g_function(b_samp, x_samp)

  # (1) Simple estimates of p and g
  p_mean <- mean(p_vals)
  g_mean <- mean(g_vals)

  # (2) Optimal gamma via covariance
  cov_pg <- cov(p_vals, g_vals)
}

```

```

var_g <- var(g_vals)
gamma_opt <- cov_pg / var_g

# (3) Construct CV estimator
p_cv <- p_mean - gamma_opt * (g_mean - Eg_true)

# This returns the CV estimate, gamma, and other useful metrics
return(list(est = p_cv))
}

```

#### 4. Importance Sampling (IS)

```

# 4.1 Define the auxiliary distribution h(b,x)
# Here we show an example of independent sampling from h(b)*h(x).
# We'll pick t_3 for b (heavier tail) and tweak Gamma parameters for x.

# b ~ t_3
r_b_IS <- function(n) {
  rt(n, df = 3)
}
# PDF of b for t_3
d_b_IS <- function(b) {
  dt(b, df = 3)
}

# x ~ Gamma(shape=2, rate=0.5)
r_x_IS <- function(n) {
  rgamma(n, shape = 2, rate = 0.5)
}
# PDF of x for Gamma(shape=2, rate=0.5)
d_x_IS <- function(x) {
  dgamma(x, shape = 2, rate = 0.5)
}

# 4.2 Implement Importance Sampling
IS_MC <- function(N) {
  # Sample from h(b,x)
  b_samp <- r_b_IS(N)
  x_samp <- r_x_IS(N)

  # Compute original PDF values f(b), f(x)
  fb_vals <- dt(b_samp, df = 5) # original t_5 for b
  fx_vals <- dgamma(x_samp, shape = 2, rate = 1) # original Gamma for x

  # Compute auxiliary PDF values h(b), h(x)
  hb_vals <- d_b_IS(b_samp)
  hx_vals <- d_x_IS(x_samp)

  # Weights w_i = [f(b_i)*f(x_i)] / [h(b_i)*h(x_i)]
  w <- (fb_vals * fx_vals) / (hb_vals * hx_vals)

  # Target function p(b,x)
  p_vals <- logistic(alpha + b_samp + beta * x_samp)
}

```

```

# IS estimate (unbiased if we average p*w)
est <- mean(p_vals * w)

# Rough variance estimate
#p_var <- mean((p_vals * w)^2) - p_hat^2
#se <- sqrt(p_var / N)

#return(list(est = est, se = se, w_mean = mean(w)))
return(list(est = est))
}

```

## Results

### 5. Comparing Bias, Variance, and CPU Time for Simple MC, CV, and Importance Sampling

```

# 5.1. Obtain a "true value" via a large-scale Simple MC
set.seed(9999)
N_large <- 5e6

res_truth <- simpleMC(N_large)
true_value <- res_truth$est
# cat("Approx. True Value =", true_value, "\n")

# 5.2. Repeat simulations for smaller N and compare bias, variance
simulate_three_methods <- function(N, Eg_true) {
  # Simple MC
  res_mc <- simpleMC(N)
  # CV
  res_cv <- cvMC(N, Eg_true = Eg_true)
  # IS
  res_is <- IS_MC(N)

  c(mc = res_mc$est, cv = res_cv$est, is = res_is$est)
}

set.seed(1234)
N_reps <- 50          # number of repeated simulations
N_small <- 1e5        # sample size for each repeated run

# Pre-calculate Eg_true for CV (to avoid doing it multiple times)
Eg_val <- approx_Eg(N=5e5)

results_matrix <- replicate(N_reps, simulate_three_methods(N_small, Eg_true = Eg_val))
# results_matrix is now 3 x N_reps, each column is one replicate

mc_est <- results_matrix["mc", ]
cv_est <- results_matrix["cv", ]
is_est <- results_matrix["is", ]

# Compute mean estimate, bias, variance

```

```

mc_mean <- mean(mc_est)
cv_mean <- mean(cv_est)
is_mean <- mean(is_est)

mc_bias <- mc_mean - true_value
cv_bias <- cv_mean - true_value
is_bias <- is_mean - true_value

mc_var <- var(mc_est)
cv_var <- var(cv_est)
is_var <- var(is_est)

df_comparison <- data.frame(
  method = c("MC", "CV", "IS", "True"),
  mean_est = c(mc_mean, cv_mean, is_mean, true_value),
  bias = c(mc_bias, cv_bias, is_bias, NA),
  variance = c(mc_var, cv_var, is_var, NA)
)

# 5.3 Compare CPU time
test_time <- microbenchmark(
  MC = { simpleMC(N_small) },
  CV = { cvMC(N_small, Eg_true = Eg_val) },
  IS = { IS_MC(N_small) },
  times = 10 # repeat each method 10 times
)

```

Table 1: Estimated model mean, bias, and variance for each method using 100,000 samples; ‘True’ value is the MC result with 6,000,000 samples

Method	Estimated Mean	Bias	Variance
MC	0.7857719	-0.000144669	4.420e-07
CV	0.7857767	-0.000139931	8.000e-09
IS	0.7855314	-0.000385163	3.016e-06
True	0.7859166	NA	NA

Table 2: CPU time statistics (ms) for each method

Method	Min.	Lower Quartile	Mean	Median	Upper Quartile	Max.	# Eval
MC	13.72758	13.86915	13.99927	14.00097	14.05886	14.34889	10
CV	16.63850	16.65084	16.86150	16.91703	16.99626	17.06584	10
IS	34.85221	35.14815	36.51008	35.43017	35.82031	41.48388	10

**Monte Carlo (MC)** is the fastest because it only generates random samples and computes the mean, with  $O(N)$  complexity and no additional calculations.

**Control Variates (CV)** is slightly slower due to the extra step of computing a regression coefficient to reduce variance, but it improves estimation accuracy.

**Importance Sampling (IS)** is the slowest because it requires computing importance weights, normalizing them, and handling probability density functions, making it computationally more expensive ( $O(N \log N)$  or higher).

## 6. Cumulative Convergence Plots (Extra Credit)

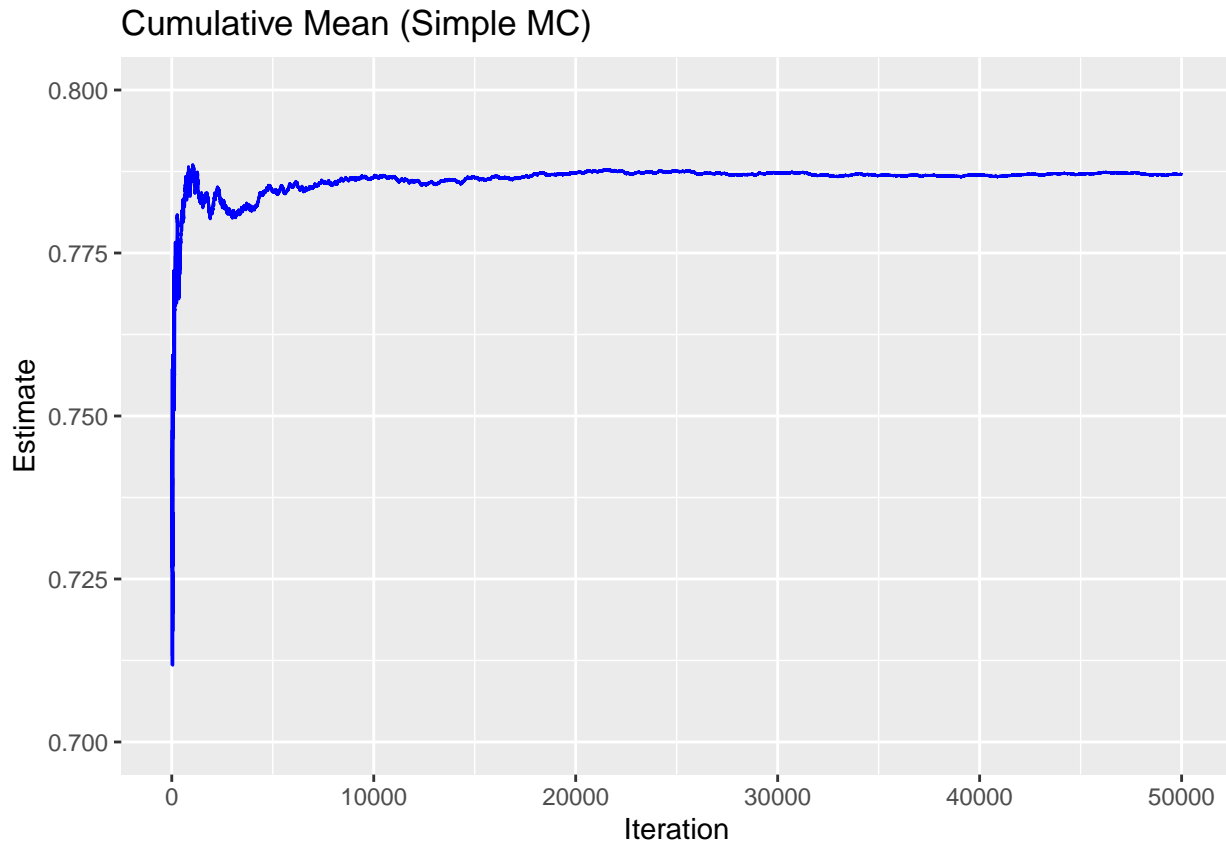
### Simple MC

```
# 6.1 Example of cumulative mean for Simple MC
cumulative_MC <- function(N) {
  b_samp <- r_b(N)
  x_samp <- r_x(N)
  p_vals <- logistic(alpha + b_samp + beta * x_samp)
  cum_mean <- cumsum(p_vals) / seq_len(N)
  return(cum_mean)
}

# 6.2 Plot cumulative mean
set.seed(999)
N <- 5e4
cm_mc <- cumulative_MC(N)

df_mc <- data.frame(
  iter = 1:N,
  mean_est = cm_mc
)

ggplot(df_mc, aes(x = iter, y = mean_est)) +
  geom_line(color = "blue") +
  labs(title = "Cumulative Mean (Simple MC)",
       x = "Iteration",
       y = "Estimate") +
  ylim(c(0.7, 0.8))
```



### Control Variate Cumulative Mean

```
# This function returns a vector of CV estimates for i from 1 to N.
cumulative_CV <- function(N, Eg_true = NULL, largeN_approx = 5e5) {
  if (is.null(Eg_true)) {
    # If E[g] is not provided, approximate with a larger MC
    Eg_true <- approx_Eg(N = largeN_approx)
  }

  # Sample from original distributions
  b_samp <- r_b(N)
  x_samp <- r_x(N)

  # Compute p(b,x) and g(b,x)
  p_vals <- logistic(alpha + b_samp + beta * x_samp)
  g_vals <- g_function(b_samp, x_samp)

  # We will keep track of partial sums to compute means, covariance, etc.
  cum_p <- cumsum(p_vals)
  cum_g <- cumsum(g_vals)
  cum_pg <- cumsum(p_vals * g_vals) # needed for Cov(p,g)
  cum_g2 <- cumsum(g_vals^2) # needed for Var(g)

  # Vector to store CV estimates at each iteration
  pCV_cum <- numeric(N)
```

```

# The first data point (i=1) does not allow us to compute a covariance
# We can simply set pCV_cum[1] to p_vals[1], or NA, etc.
pCV_cum[1] <- p_vals[1]

# Loop over i = 2 to N
for (i in 2:N) {
  # partial means
  p_bar_i <- cum_p[i] / i
  g_bar_i <- cum_g[i] / i

  # Cov(p,g) = E[ p*g ] - E[p]*E[g ]
  # partial estimate of E[p*g] = cum_pg[i] / i
  pg_bar_i <- cum_pg[i] / i
  cov_pg_i <- pg_bar_i - (p_bar_i * g_bar_i)

  # Var(g) = E[g^2] - (E[g])^2
  g2_bar_i <- cum_g2[i] / i
  var_g_i <- g2_bar_i - (g_bar_i^2)

  # Optimal gamma
  gamma_i <- cov_pg_i / var_g_i

  # CV estimate at iteration i
  pCV_cum[i] <- p_bar_i - gamma_i * (g_bar_i - Eg_true)
}

return(pCV_cum)
}

# Example usage and plotting:
set.seed(1234)
N <- 20000
cv_cum_values <- cumulative_CV(N)

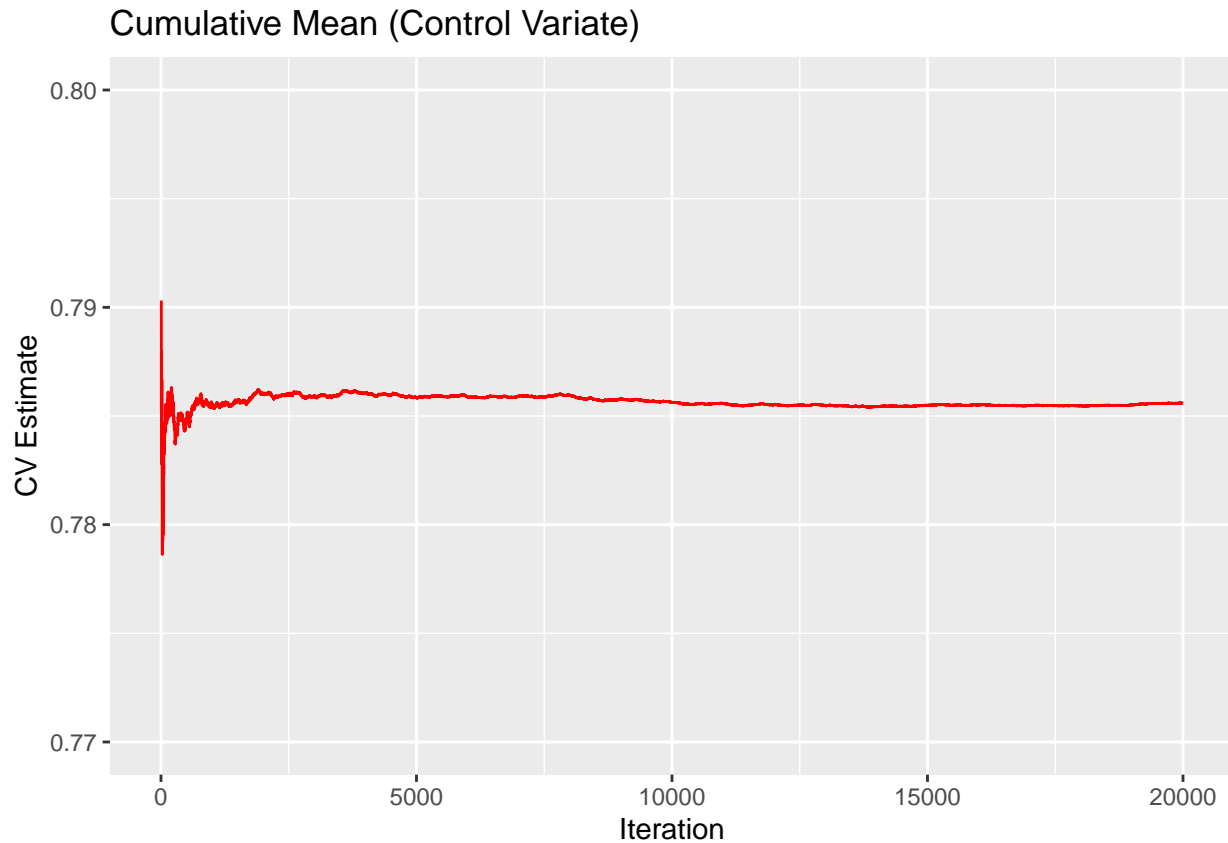
df_cv <- data.frame(
  iter = 1:N,
  cv_est = cv_cum_values
)

library(ggplot2)
ggplot(df_cv, aes(x = iter, y = cv_est)) +
  geom_line(color = "red") +
  labs(title = "Cumulative Mean (Control Variate)",
       x = "Iteration",
       y = "CV Estimate") +
  ylim(c(0.77, 0.8))

## Warning: Removed 5 rows containing missing values or values outside the scale range
## (`geom_line()`).

```





### Importance Sampling Cumulative Mean

```
cumulative_IS <- function(N) {
  # Sample from auxiliary distribution h(b,x)
  b_samp <- r_b_IS(N)
  x_samp <- r_x_IS(N)

  # Compute original PDF f(b), f(x)
  fb_vals <- dt(b_samp, df = 5)          # original b ~ t_5
  fx_vals <- dgamma(x_samp, shape=2, rate=1) # original x ~ Gamma(2,1)

  # Compute auxiliary PDF h(b), h(x)
  hb_vals <- d_b_IS(b_samp)
  hx_vals <- d_x_IS(x_samp)

  # Weights
  w <- (fb_vals * fx_vals) / (hb_vals * hx_vals)

  # Target function p(b,x)
  p_vals <- logistic(alpha + b_samp + beta * x_samp)

  # Build cumulative sums
  cum_w <- cumsum(w)
  cum_pw <- cumsum(p_vals * w)

  # IS estimate at iteration i: cum_pw[i] / cum_w[i]
}
```

```

pIS_cum <- cum_pw / cum_w

  return(pIS_cum)
}

## Example usage and plotting:
set.seed(5678)
N <- 20000
is_cum_values <- cumulative_IS(N)

df_is <- data.frame(
  iter = 1:N,
  is_est = is_cum_values
)

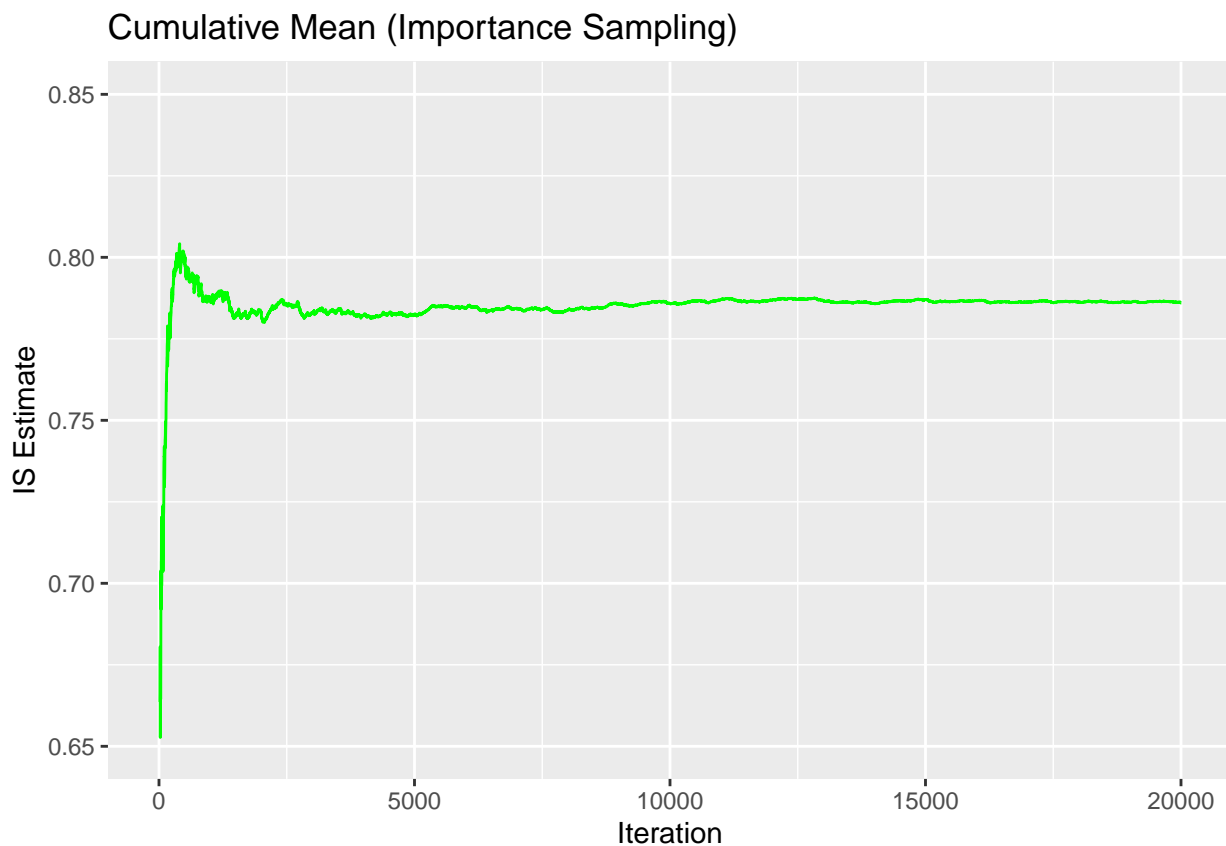
ggplot(df_is, aes(x = iter, y = is_est)) +
  geom_line(color = "green") +
  labs(title = "Cumulative Mean (Importance Sampling)",
       x = "Iteration",
       y = "IS Estimate") +
  ylim(c(0.65, 0.85))

```

```

## Warning: Removed 20 rows containing missing values or values outside the scale range
## (`geom_line()`).

```



```

N <- 20000
set.seed(999)

```

```

# 1. Simple MC
cm_mc <- cumulative_MC(N)
df_mc <- data.frame(iter = 1:N, method = "SimpleMC", estimate = cm_mc)

# 2. CV
cm_cv <- cumulative_CV(N)
df_cv <- data.frame(iter = 1:N, method = "ControlVar", estimate = cm_cv)

# 3. IS
cm_is <- cumulative_IS(N)
df_is <- data.frame(iter = 1:N, method = "ImpSampling", estimate = cm_is)

# Combine
df_all <- rbind(df_mc, df_cv, df_is)

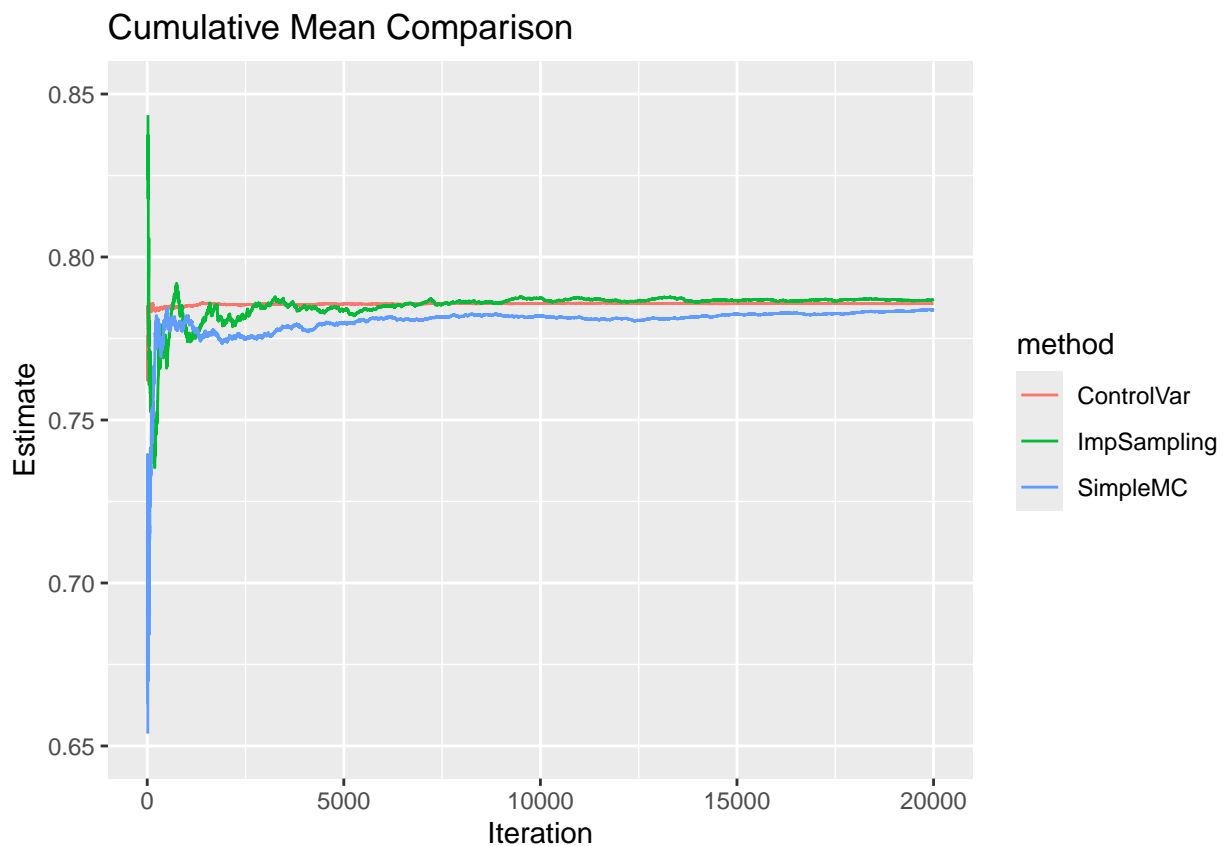
ggplot(df_all, aes(x = iter, y = estimate, color = method)) +
  geom_line() +
  labs(title = "Cumulative Mean Comparison",
       x = "Iteration",
       y = "Estimate") +
  ylim(c(0.65, 0.85))

```

```

## Warning: Removed 20 rows containing missing values or values outside the scale range
## (`geom_line()`).

```



```
theme_bw()
```

```

## List of 136
## $ line                                     :List of 6
##   ..$ colour          : chr "black"
##   ..$ linewidth       : num 0.5
##   ..$ linetype        : num 1
##   ..$ lineend         : chr "butt"
##   ..$ arrow           : logi FALSE
##   ..$ inherit.blank   : logi TRUE
##   ..- attr(*, "class")= chr [1:2] "element_line" "element"
## $ rect                                     :List of 5
##   ..$ fill            : chr "white"
##   ..$ colour          : chr "black"
##   ..$ linewidth       : num 0.5
##   ..$ linetype        : num 1
##   ..$ inherit.blank   : logi TRUE
##   ..- attr(*, "class")= chr [1:2] "element_rect" "element"
## $ text                                    :List of 11
##   ..$ family          : chr ""
##   ..$ face            : chr "plain"
##   ..$ colour          : chr "black"
##   ..$ size            : num 11
##   ..$ hjust           : num 0.5
##   ..$ vjust           : num 0.5
##   ..$ angle           : num 0
##   ..$ lineheight      : num 0.9
##   ..$ margin          : 'margin' num [1:4] 0points 0points 0points 0points
##   .. ..- attr(*, "unit")= int 8
##   ..$ debug           : logi FALSE
##   ..$ inherit.blank   : logi TRUE
##   ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ title                                     : NULL
## $ aspect.ratio                                     : NULL
## $ axis.title                                     : NULL
## $ axis.title.x                                   :List of 11
##   ..$ family          : NULL
##   ..$ face            : NULL
##   ..$ colour          : NULL
##   ..$ size            : NULL
##   ..$ hjust           : NULL
##   ..$ vjust           : num 1
##   ..$ angle           : NULL
##   ..$ lineheight      : NULL
##   ..$ margin          : 'margin' num [1:4] 2.75points 0points 0points 0points
##   .. ..- attr(*, "unit")= int 8
##   ..$ debug           : NULL
##   ..$ inherit.blank   : logi TRUE
##   ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.title.x.top                             :List of 11
##   ..$ family          : NULL
##   ..$ face            : NULL
##   ..$ colour          : NULL
##   ..$ size            : NULL
##   ..$ hjust           : NULL
##   ..$ vjust           : num 0

```

```

## ..$ angle      : NULL
## ..$ lineheight : NULL
## ..$ margin      : 'margin' num [1:4] 0points 0points 2.75points 0points
## ..- attr(*, "unit")= int 8
## ..$ debug       : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.title.x.bottom : NULL
## $ axis.title.y        :List of 11
## ..$ family           : NULL
## ..$ face              : NULL
## ..$ colour            : NULL
## ..$ size              : NULL
## ..$ hjust             : NULL
## ..$ vjust             : num 1
## ..$ angle             : num 90
## ..$ lineheight        : NULL
## ..$ margin            : 'margin' num [1:4] 0points 2.75points 0points 0points
## ..- attr(*, "unit")= int 8
## ..$ debug             : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.title.y.left   : NULL
## $ axis.title.y.right  :List of 11
## ..$ family           : NULL
## ..$ face              : NULL
## ..$ colour            : NULL
## ..$ size              : NULL
## ..$ hjust             : NULL
## ..$ vjust             : num 1
## ..$ angle             : num -90
## ..$ lineheight        : NULL
## ..$ margin            : 'margin' num [1:4] 0points 0points 0points 2.75points
## ..- attr(*, "unit")= int 8
## ..$ debug             : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.text           :List of 11
## ..$ family           : NULL
## ..$ face              : NULL
## ..$ colour            : chr "grey30"
## ..$ size              : 'rel' num 0.8
## ..$ hjust             : NULL
## ..$ vjust             : NULL
## ..$ angle             : NULL
## ..$ lineheight        : NULL
## ..$ margin            : NULL
## ..$ debug             : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.text.x         :List of 11
## ..$ family           : NULL
## ..$ face              : NULL
## ..$ colour            : NULL

```

```

## ..$ size      : NULL
## ..$ hjust     : NULL
## ..$ vjust     : num 1
## ..$ angle     : NULL
## ..$ lineheight : NULL
## ..$ margin    : 'margin' num [1:4] 2.2points 0points 0points 0points
## .. ..- attr(*, "unit")= int 8
## ..$ debug     : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.text.x.top      :List of 11
## ..$ family           : NULL
## ..$ face             : NULL
## ..$ colour          : NULL
## ..$ size            : NULL
## ..$ hjust           : NULL
## ..$ vjust           : num 0
## ..$ angle           : NULL
## ..$ lineheight      : NULL
## ..$ margin          : 'margin' num [1:4] 0points 0points 2.2points 0points
## .. ..- attr(*, "unit")= int 8
## ..$ debug           : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.text.x.bottom   : NULL
## $ axis.text.y          :List of 11
## ..$ family           : NULL
## ..$ face             : NULL
## ..$ colour          : NULL
## ..$ size            : NULL
## ..$ hjust           : num 1
## ..$ vjust           : NULL
## ..$ angle           : NULL
## ..$ lineheight      : NULL
## ..$ margin          : 'margin' num [1:4] 0points 2.2points 0points 0points
## .. ..- attr(*, "unit")= int 8
## ..$ debug           : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.text.y.left     : NULL
## $ axis.text.y.right    :List of 11
## ..$ family           : NULL
## ..$ face             : NULL
## ..$ colour          : NULL
## ..$ size            : NULL
## ..$ hjust           : num 0
## ..$ vjust           : NULL
## ..$ angle           : NULL
## ..$ lineheight      : NULL
## ..$ margin          : 'margin' num [1:4] 0points 0points 0points 2.2points
## .. ..- attr(*, "unit")= int 8
## ..$ debug           : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"

```

```

## $ axis.text.theta : NULL
## $ axis.text.r :List of 11
## ..$ family : NULL
## ..$ face : NULL
## ..$ colour : NULL
## ..$ size : NULL
## ..$ hjust : num 0.5
## ..$ vjust : NULL
## ..$ angle : NULL
## ..$ lineheight : NULL
## ..$ margin : 'margin' num [1:4] 0points 2.2points 0points 2.2points
## ..- attr(*, "unit")= int 8
## ..$ debug : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ axis.ticks :List of 6
## ..$ colour : chr "grey20"
## ..$ linewidth : NULL
## ..$ linetype : NULL
## ..$ lineend : NULL
## ..$ arrow : logi FALSE
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_line" "element"
## $ axis.ticks.x : NULL
## $ axis.ticks.x.top : NULL
## $ axis.ticks.x.bottom : NULL
## $ axis.ticks.y : NULL
## $ axis.ticks.y.left : NULL
## $ axis.ticks.y.right : NULL
## $ axis.ticks.theta : NULL
## $ axis.ticks.r : NULL
## $ axis.minor.ticks.x.top : NULL
## $ axis.minor.ticks.x.bottom : NULL
## $ axis.minor.ticks.y.left : NULL
## $ axis.minor.ticks.y.right : NULL
## $ axis.minor.ticks.theta : NULL
## $ axis.minor.ticks.r : NULL
## $ axis.ticks.length : 'simpleUnit' num 2.75points
## ..- attr(*, "unit")= int 8
## $ axis.ticks.length.x : NULL
## $ axis.ticks.length.x.top : NULL
## $ axis.ticks.length.x.bottom : NULL
## $ axis.ticks.length.y : NULL
## $ axis.ticks.length.y.left : NULL
## $ axis.ticks.length.y.right : NULL
## $ axis.ticks.length.theta : NULL
## $ axis.ticks.length.r : NULL
## $ axis.minor.ticks.length : 'rel' num 0.75
## $ axis.minor.ticks.length.x : NULL
## $ axis.minor.ticks.length.x.top : NULL
## $ axis.minor.ticks.length.x.bottom: NULL
## $ axis.minor.ticks.length.y : NULL
## $ axis.minor.ticks.length.y.left : NULL
## $ axis.minor.ticks.length.y.right : NULL

```

```

## $ axis.minor.ticks.length.theta : NULL
## $ axis.minor.ticks.length.r : NULL
## $ axis.line : list()
## ..- attr(*, "class")= chr [1:2] "element_blank" "element"
## $ axis.line.x : NULL
## $ axis.line.x.top : NULL
## $ axis.line.x.bottom : NULL
## $ axis.line.y : NULL
## $ axis.line.y.left : NULL
## $ axis.line.y.right : NULL
## $ axis.line.theta : NULL
## $ axis.line.r : NULL
## $ legend.background :List of 5
## ..$ fill : NULL
## ..$ colour : logi NA
## ..$ linewidth : NULL
## ..$ linetype : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_rect" "element"
## $ legend.margin : 'margin' num [1:4] 5.5points 5.5points 5.5points 5.5points
## ..- attr(*, "unit")= int 8
## $ legend.spacing : 'simpleUnit' num 11points
## ..- attr(*, "unit")= int 8
## $ legend.spacing.x : NULL
## $ legend.spacing.y : NULL
## $ legend.key : NULL
## $ legend.key.size : 'simpleUnit' num 1.2lines
## ..- attr(*, "unit")= int 3
## $ legend.key.height : NULL
## $ legend.key.width : NULL
## $ legend.key.spacing : 'simpleUnit' num 5.5points
## ..- attr(*, "unit")= int 8
## $ legend.key.spacing.x : NULL
## $ legend.key.spacing.y : NULL
## $ legend.frame : NULL
## $ legend.ticks : NULL
## $ legend.ticks.length : 'rel' num 0.2
## $ legend.axis.line : NULL
## $ legend.text :List of 11
## ..$ family : NULL
## ..$ face : NULL
## ..$ colour : NULL
## ..$ size : 'rel' num 0.8
## ..$ hjust : NULL
## ..$ vjust : NULL
## ..$ angle : NULL
## ..$ lineheight : NULL
## ..$ margin : NULL
## ..$ debug : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ legend.text.position : NULL
## $ legend.title :List of 11
## ..$ family : NULL

```



```

## ..$ face          : NULL
## ..$ colour        : NULL
## ..$ size          : NULL
## ..$ hjust         : num 0
## ..$ vjust         : NULL
## ..$ angle         : NULL
## ..$ lineheight    : NULL
## ..$ margin        : NULL
## ..$ debug         : NULL
## ..$ inherit.blank: logi TRUE
## ..- attr(*, "class")= chr [1:2] "element_text" "element"
## $ legend.title.position : NULL
## $ legend.position       : chr "right"
## $ legend.position.inside : NULL
## $ legend.direction      : NULL
## $ legend.byrow          : NULL
## $ legend.justification  : chr "center"
## $ legend.justification.top : NULL
## $ legend.justification.bottom : NULL
## $ legend.justification.left : NULL
## $ legend.justification.right : NULL
## $ legend.justification.inside : NULL
## $ legend.location       : NULL
## $ legend.box            : NULL
## $ legend.box.just       : NULL
## $ legend.box.margin     : 'margin' num [1:4] 0cm 0cm 0cm 0cm
## ..- attr(*, "unit")= int 1
## $ legend.box.background : list()
## ..- attr(*, "class")= chr [1:2] "element_blank" "element"
## $ legend.box.spacing    : 'simpleUnit' num 11points
## ..- attr(*, "unit")= int 8
## [list output truncated]
## - attr(*, "class")= chr [1:2] "theme" "gg"
## - attr(*, "complete")= logi TRUE
## - attr(*, "validate")= logi TRUE

```

## Discussion and Practical Implications

## Conclusion