## Question1\_Rmd

## clear working memory

```
rm(list=ls())
loading libraries
## Attaching package: 'rmutil'
## The following object is masked from 'package:stats':
##
##
      nobs
## The following objects are masked from 'package:base':
##
##
      as.data.frame, units
## Registered S3 method overwritten by 'httr':
##
    method
                  from
    print.response rmutil
## -- Attaching packages ------ tidyverse 1.3.1 --
## v ggplot2 3.3.5
                               0.3.4
                   v purrr
## v tibble 3.1.4
                    v dplyr
                              1.0.7
          1.1.3
                     v stringr 1.4.0
## v tidyr
## v readr
          2.0.1
                     v forcats 0.5.1
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                  masks stats::lag()
## x tidyr::nesting() masks rmutil::nesting()
# For this exercise, you will use a simulation to see how well the CLT works with finite samples in R.
#a) Suppose that x is binary with Pr(x = 1) = 0.35.
generate_simulated_means <- function(N){</pre>
   \textit{\# Generate the mean and standard deviations of N observations from the specified distribution functi} \\
```

# x is binary with Pr(x = 1) = 0.35

```
\# runif(n) returns a vector (length n) of random draw from the uniform distribution U[0,1]
   x \leftarrow ifelse(runif(N) < 0.35, 1, 0)
   # put data into data_frame so it is easier to summarize
   data <- tibble(x)</pre>
   # get the means for each column
   means <- sapply(data, mean)</pre>
   # name the means appropriately
   names(means) <- c("mu1")</pre>
   # get the sds for each column
   sds <- sapply(data, sd)</pre>
   # name the sds appropriately
   names(sds) \leftarrow c("sd1")
   # return the means and standard deviation associated with sample x of size N.
   return(c(means, sds))
   }
# Finding the simulated means and sd for our distributions
# with sample sizes 36, 64, 100, 225, 2500, and 12100
# for each sample size 10,000 replications.
# Using CLT to see how far our observed means were from
# the true means of each distribution.
# Calculating z-scores and then see empirically how many of the means
# were beyond our critical values.
get_zscores <-function(obs_mean, true_mean, obs_sd, N){</pre>
  zscores <- (obs_mean - true_mean) / (obs_sd / sqrt(N))</pre>
  return( zscores )
significance_test <- function(zscores, alpha){</pre>
  beyond_critical_point <- as.numeric( zscores > alpha | zscores < -alpha )</pre>
  percent_significantly_different <- mean( beyond_critical_point )</pre>
  return( percent_significantly_different )
}
monte_carlo <- function(N, reps = 10000){</pre>
  replicated_sims <- replicate(reps, generate_simulated_means(N))</pre>
  expected_mu <- 0.35
```

```
z1 <- get_zscores(replicated_sims['mu1', ], expected_mu, replicated_sims['sd1', ], N)
  sig1 <- significance_test(z1, 0.025)</pre>
  print(paste("Percentage of simulated means which were significantly different from"))
  print(paste("sampling distribution at critical point 0.025:", sig1))
  print(paste("
  sig2 <- significance test(z1, 0.975)</pre>
  print(paste("Percentage of simulated means which were significantly different from"))
  print(paste("sampling distribution at critical point 0.975:", sig2))
  print(paste("
}
for (N in c(36, 64, 100, 225, 2500, 12100)){
  print(paste('Starting simulations with samples of size', N))
  monte_carlo(N, 10000)
  print('')
}
## [1] "Starting simulations with samples of size 36"
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.025: 1"
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.975: 0.2918"
## [1] "
## [1] ""
## [1] "Starting simulations with samples of size 64"
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.025: 1"
## [1] "
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.975: 0.2935"
## [1] "
## [1] ""
## [1] "Starting simulations with samples of size 100"
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.025: 0.9143"
## [1] "
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.975: 0.3452"
## [1] "
## [1] ""
## [1] "Starting simulations with samples of size 225"
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.025: 1"
## [1] "
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.975: 0.3328"
## [1] "
## [1] ""
```

```
## [1] "Starting simulations with samples of size 2500"
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.025: 0.9806"
## [1] "
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.975: 0.3223"
## [1] "
## [1] ""
## [1] "Starting simulations with samples of size 12100"
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.025: 0.9776"
## [1] "
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.975: 0.3247"
## [1] "
## [1] ""
#1b) Suppose that x is binary with Pr(x = 1) = 0.97.
generate_simulated_means <- function(N){</pre>
   # Generate the mean and standard deviations of N observations from the specified distribution functi
   # x is binary with Pr(x = 1) = 0.97
   # runif(n) returns a vector (length n) of random draw from the uniform distribution U[0,1]
  x \leftarrow ifelse(runif(N) < .97, 1, 0)
   # put data into data_frame so it is easier to summarize
   data <- data_frame(x)</pre>
   # get the means for each column
  means <- sapply(data, mean)</pre>
   # name the means appropriately
   names(means) <- c("mu1")</pre>
   # get the sds for each column
   sds <- sapply(data, sd)</pre>
   # name the sds appropriately
  names(sds) \leftarrow c("sd1")
   # return the means and standard deviation associated with sample x of size N.
   return(c(means, sds))
   }
# Finding the simulated means and sd for our distributions
# with sample sizes 36, 64, 100, 225, 2500, and 12100
# for each sample size 10,000 replications.
# Using CLT to see how far our observed means were from
# the true means of each distribution.
# Calculating z-scores and then see empirically how many of the means
# were beyond our critical values.
```

```
get_zscores <-function(obs_mean, true_mean, obs_sd, N){</pre>
  zscores <- (obs_mean - true_mean) / (obs_sd / sqrt(N))</pre>
 return( zscores )
significance_test <- function(zscores, alpha){</pre>
  beyond_critical_point <- as.numeric( zscores > alpha | zscores < -alpha )</pre>
 percent significantly different <- mean( beyond critical point )</pre>
 return( percent_significantly_different )
monte_carlo <- function(N, reps = 10000){</pre>
  replicated_sims <- replicate(reps, generate_simulated_means(N))</pre>
  expected_mu <- 0.97
 z1 <- get_zscores(replicated_sims['mu1', ], expected_mu, replicated_sims['sd1', ], N)
  sig1 <- significance_test(z1, 0.025)</pre>
  print(paste("Percentage of simulated means which were significantly different from"))
  print(paste("sampling distribution at critical point 0.025:", sig1))
  print(paste("
  sig2 <- significance_test(z1, 0.975)</pre>
  print(paste("Percentage of simulated means which were significantly different from"))
  print(paste("sampling distribution at critical point 0.975:", sig2))
  print(paste("
                                    "))
}
for (N in c(36, 64, 100, 225, 2500, 12100)){
  print(paste('Starting simulations with samples of size', N))
 monte carlo(N, 10000)
 print('')
}
## [1] "Starting simulations with samples of size 36"
## Warning: 'data_frame()' was deprecated in tibble 1.1.0.
## Please use 'tibble()' instead.
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.025: 1"
## [1] "
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.975: 0.4349"
## [1] "
## [1] ""
```

```
## [1] "Starting simulations with samples of size 64"
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.025: 1"
## [1] "
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.975: 0.2704"
## [1] "
## [1] ""
## [1] "Starting simulations with samples of size 100"
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.025: 0.7758"
## [1] "
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.975: 0.2772"
## [1] "
## [1] ""
## [1] "Starting simulations with samples of size 225"
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.025: 1"
## [1] "
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.975: 0.3457"
## [1] "
## [1] ""
## [1] "Starting simulations with samples of size 2500"
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.025: 0.9512"
## [1] "
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.975: 0.3507"
## [1] "
## [1] ""
## [1] "Starting simulations with samples of size 12100"
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.025: 0.9787"
## [1] "
## [1] "Percentage of simulated means which were significantly different from"
## [1] "sampling distribution at critical point 0.975: 0.3479"
## [1] "
## [1] ""
```

## Observations

Central Limit theorem is interpreted here. 1. As we increase N, the percentage of sample means that have a z-score below -0.025 and above 0.025 is  $\sim$ 99%. 2. For critical point z=0.975: As we increase N, the percentage of sample means that have a z-score below -0.975 and above 0.975 is  $\sim$ 34%, which means 66% of the sample means are between z score of 0.975. These simulation results are in accordance with a typical normal distribution where almost 68% of sample means lie within a z-score of 1 and where many sample means fall outside the z-score of 0.025 as the interval defined by the same is very very small.