## HW2 Solution

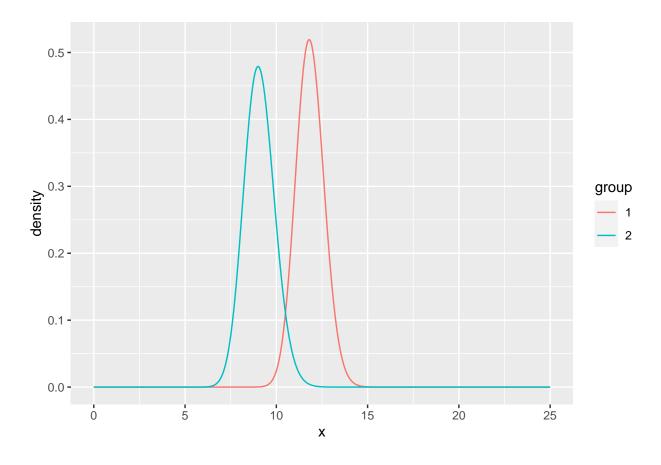
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## Question 2

**a**)

```
library(tidyverse)
## -- Attaching packages -----
                                           ----- tidyverse 1.3.1 --
## v ggplot2 3.4.0
                      v purrr
                                0.3.4
## v tibble 3.1.6
                      v dplyr
                                1.0.7
## v tidyr 1.1.4 v stringr 1.4.0
## v readr
           2.1.1
                      v forcats 0.5.1
## -- Conflicts -----
                                        ----- tidyverse conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                    masks stats::lag()
params \leftarrow tibble(alpha = c(237, 118), beta = c(20, 13))
library(HDInterval)
for (i in 1:2) {
 print(hdi(qgamma, 0.95, shape = params$alpha[i], rate = params$beta[i]))
     lower
              upper
## 10.35737 13.37075
## attr(,"credMass")
## [1] 0.95
      lower
                upper
## 7.465163 10.732044
## attr(,"credMass")
## [1] 0.95
n_grid <- 1000
plot_tb <- bind_rows(</pre>
 tibble(x = seq(from = 0, to = 25, length.out = n_grid),
                 density = dgamma(x, shape = params$alpha[1], rate = params$beta[1]),
                 group = '1'),
 tibble(x = seq(from = 0, to = 25, length.out = n_grid),
                 density = dgamma(x, shape = params$alpha[2], rate = params$beta[2]),
                 group = '2')
plot_tb %>%
 ggplot(aes(x = x, y = density, color = group)) +
 geom_line()
```



```
b)
B <- 50
xs <- 1:B
sum(dnbinom(xs, size = params$alpha[1], prob = params$beta[1] / (params$beta[1] + 1)) *
      pnbinom(xs - 1, size = params$alpha[2], prob = params$beta[2] / (params$beta[2] + 1)))
## [1] 0.6856978
xs <- 0:B
sum(dnbinom(xs, size = params$alpha[1], prob = params$beta[1] / (params$beta[1] + 1)) *
      pnbinom(xs, size = params$alpha[2], prob = params$beta[2] / (params$beta[2] + 1)))
## [1] 0.7575315
c)
With 95% probability, the range of asymptotic confidence interval is 2*z_{0.975}\sqrt{\frac{\sigma^2}{n}} which should equals 0.05 *
2.
variance <- sum(params$alpha / params$beta^2)</pre>
variance / 0.05^2 * qnorm(0.975)^2
## [1] 1983.307
```

d)

```
set.seed(0)
n <- 10000
theta_as <- rgamma(n, params$alpha[1], params$beta[1])
theta_bs <- rgamma(n, params$alpha[2], params$beta[2])
mean(theta_as - theta_bs)

## [1] 2.768717
mean(theta_as > theta_bs)

## [1] 0.9911
quantile(theta_as - theta_bs, probs = c(0.025, 0.975))

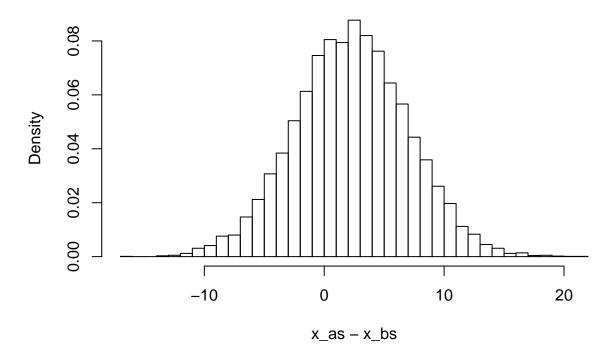
## 2.5% 97.5%
## 0.4943001 4.9350317
```

From the confidence interval, we know  $\theta_a$  is significantly larger than  $\theta_b$ , which is consistent with the  $P(\theta_a > \theta_b)$  estimation.

**e**)

```
set.seed(0)
n <- 10000
x_as <- rnbinom(n, size = params$alpha[1], prob = params$beta[1] / (params$beta[1] + 1))
x_bs <- rnbinom(n, size = params$alpha[2], prob = params$beta[2] / (params$beta[2] + 1))
hist(x_as - x_bs, freq = FALSE, breaks = 50)</pre>
```

## Histogram of x\_as - x\_bs



The result is supposed to be the difference between two negative binomial random variables.

## Question 4

```
set.seed(0)
n <- 10000
xs <- runif(n)
mean(4 * sqrt(1 - xs^2))

## [1] 3.134347
set.seed(0)
n <- 10000
xs <- rbeta(n, 2, 2)
mean(4 * sqrt(1 - xs^2) / dbeta(xs, 2, 2))

## [1] 3.125043</pre>
```