bios6301hw9

Haoyang Yi

11/9/2022

Question 1

15 points

Consider the following very simple genetic model (*very* simple – don't worry if you're not a geneticist!). A population consists of equal numbers of two sexes: male and female. At each generation men and women are paired at random, and each pair produces exactly two offspring, one male and one female. We are interested in the distribution of height from one generation to the next. Suppose that the height of both children is just the average of the height of their parents, how will the distribution of height change across generations?

Represent the heights of the current generation as a dataframe with two variables, m and f, for the two sexes. We can use **rnorm** to randomly generate the population at generation 1:

```
pop <- data.frame(m = rnorm(100, 160, 20), f = rnorm(100, 160, 20))</pre>
```

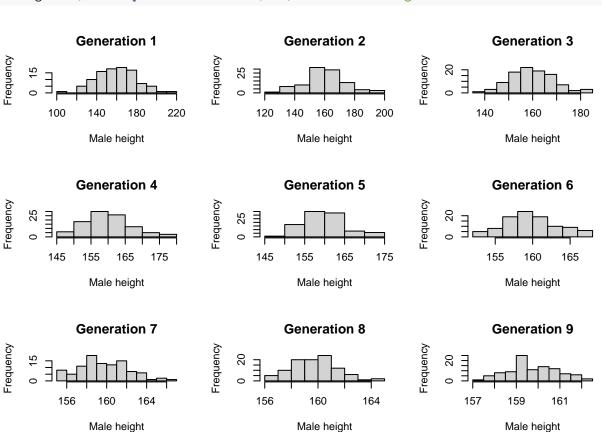
The following function takes the data frame pop and randomly permutes the ordering of the men. Men and women are then paired according to rows, and heights for the next generation are calculated by taking the mean of each row. The function returns a data frame with the same structure, giving the heights of the next generation.

```
next_gen <- function(pop) {
   pop$m <- sample(pop$m)
   pop$m <- rowMeans(pop)
   pop$f <- pop$m
   pop
}</pre>
```

Use the function <code>next_gen</code> to generate nine generations (you already have the first), then use the function <code>hist</code> to plot the distribution of male heights in each generation (this will require multiple calls to <code>hist</code>). The phenomenon you see is called regression to the mean. Provide (at least) minimal decorations such as title and x-axis labels.

```
gen1 = pop
gen2 = next_gen(gen1)
gen3 = next_gen(gen2)
gen4 = next_gen(gen3)
gen5 = next_gen(gen4)
gen6 = next_gen(gen5)
gen7 = next_gen(gen6)
gen8 = next_gen(gen7)
gen9 = next_gen(gen8)
par(mfrow = c(3,3))
hist(gen1$m,main = paste("Generation", 1),xlab = "Male height")
hist(gen2$m,main = paste("Generation", 2),xlab = "Male height")
hist(gen3$m,main = paste("Generation", 3),xlab = "Male height")
hist(gen4$m,main = paste("Generation", 4),xlab = "Male height")
```

```
hist(gen5$m,main = paste("Generation", 5),xlab = "Male height")
hist(gen6$m,main = paste("Generation", 6),xlab = "Male height")
hist(gen7$m,main = paste("Generation", 7),xlab = "Male height")
hist(gen8$m,main = paste("Generation", 8),xlab = "Male height")
hist(gen9$m,main = paste("Generation", 9),xlab = "Male height")
```

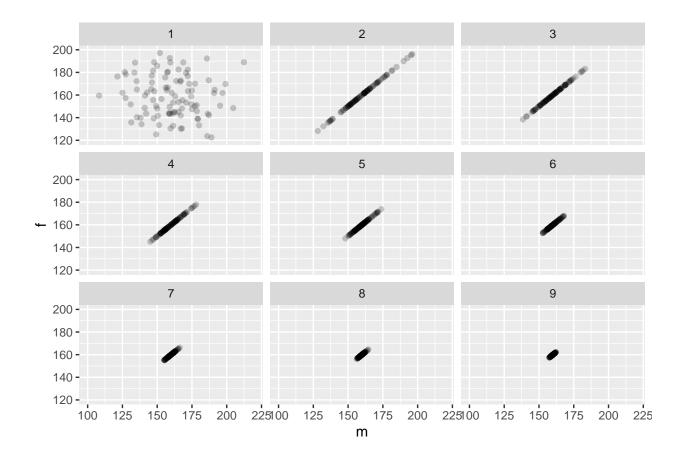


Question 2

10 points

Use the simulated results from question 1 to reproduce (as closely as possible) the following plot in ggplot2.

Warning: Removed 3 rows containing missing values (geom_point).



Question 3

15 points

You calculated the power of a study design in question #1 of assignment 3. The study has two variables, treatment group and outcome. There are two treatment groups (0, 1) and they should be assigned randomly with equal probability. The outcome should be a random normal variable with a mean of 60 and standard deviation of 20. If a patient is in the treatment group, add 5 to the outcome.

Starting with a sample size of 250, create a 95% bootstrap percentile interval for the mean of each group. Then create a new bootstrap interval by increasing the sample size by 250 until the sample is 2500. Thus you will create a total of 10 bootstrap intervals. Each bootstrap should create 1000 bootstrap samples. (9 points)

```
intervals[i,1]=n
intervals [i,2:3] = quantile (res[,1],c(0.025,0.975))
intervals[i,4:5] = quantile(res[,2],c(0.025,0.975))
intervals[i,6] = mean(res[,1])
intervals[i,7] = mean(res[,2])
intervals
      size 2.5%_treatment0 97.5%_treament0 2.5%_treatment1 97.5%_treatment1
##
## 1
       250
                  53.99656
                                   60.98433
                                                   59.12599
                                                                     65.06593
## 2
       500
                                   64.18093
                                                   64.47157
                  59.27101
                                                                     69.05103
## 3
      750
                  58.02260
                                   62.12314
                                                   63.61589
                                                                     67.82770
## 4
     1000
                  57.73422
                                   61.19670
                                                   63.23148
                                                                     66.94054
## 5
     1250
                                   59.69581
                                                                     66.68974
                  56.70952
                                                   63.81686
## 6
     1500
                  59.73296
                                   62.74158
                                                   62.71206
                                                                     65.48744
## 7 1750
                  58.69097
                                   61.35978
                                                   62.46932
                                                                     65.08222
## 8 2000
                  59.81898
                                   62.19992
                                                   64.12481
                                                                     66.55308
## 9
     2250
                  60.15856
                                   62.54635
                                                   63.00229
                                                                     65.36832
## 10 2500
                  57.68994
                                   59.81357
                                                   63.61398
                                                                     65.68806
##
       mean_0
                 mean_1
## 1 57.66908 62.08083
## 2
     61.75156 66.79496
## 3 60.07992 65.72524
## 4 59.55441 65.14387
## 5 58.17597 65.28930
     61.30692 64.10961
## 6
## 7 59.98083 63.71939
## 8 60.98619 65.30477
## 9 61.34470 64.11646
```

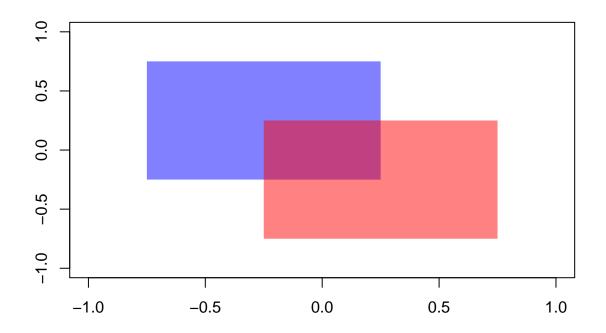
Produce a line chart that includes the bootstrapped mean and lower and upper percentile intervals for each group. Add appropriate labels and a legend. (6 points)

Here's an example of how you could create transparent shaded areas.

10 58.72048 64.67747

```
makeTransparent = function(..., alpha=0.5) {
  if(alpha<0 | alpha>1) stop("alpha must be between 0 and 1")
  alpha = floor(255*alpha)
  newColor = col2rgb(col=unlist(list(...)), alpha=FALSE)
  .makeTransparent = function(col, alpha) {
    rgb(red=col[1], green=col[2], blue=col[3], alpha=alpha, maxColorValue=255)
  }
  newColor = apply(newColor, 2, .makeTransparent, alpha=alpha)
  return(newColor)
}
par(new=FALSE)
plot(NULL,
 xlim=c(-1, 1),
  ylim=c(-1, 1),
 xlab="",
 ylab=""
polygon(x=c(seq(-0.75, 0.25, length.out=100), seq(0.25, -0.75, length.out=100)),
```

```
y=c(rep(-0.25, 100), rep(0.75, 100)), border=NA, col=makeTransparent('blue',alpha=0.5))
polygon(x=c(seq(-0.25, 0.75, length.out=100), seq(0.75, -0.25, length.out=100)),
    y=c(rep(-0.75, 100), rep(0.25, 100)), border=NA, col=makeTransparent('red',alpha=0.5))
```



```
plot(NULL,
    xlim=c(1,10),
    ylim=c(50,75),
    xlab="n-th bootstrap interval",
    ylab="outcome"
)
lines(x=seq(1:10), y = intervals$mean_1, col = "red")
polygon(x=c(seq(1,10), seq(10,1)), y = c(intervals$^2.5%_treatment1^, rev(intervals$^97.5%_treatment1^)
border=NA, col=makeTransparent('red',alpha=0.5))
lines(x=seq(1:10), y = intervals$mean_0, col = "blue")
polygon(x=c(seq(1,10), seq(10,1)), y = c(intervals$^2.5%_treatment0^, rev(intervals$^97.5%_treament0^))
border=NA, col=makeTransparent('blue',alpha=0.5))
lab <- c('treatment_1', 'treatment_0')
legend("bottomright", lab, pch=c(15,15), col=c('red','blue'))</pre>
```

