

Math 181B: Homework 6

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Exercise 1

$$\begin{aligned} E[SSTOT] &= E[SSTR + SSE] \\ &= E[SSTR] + E[SSE] \\ &= E[SSTR] + E\left[\sum_{j=1}^k (n_j - 1)S_j^2\right] \\ &= E[SSTR] + \sum_{j=1}^k (n_j - 1)\sigma^2 \\ &= E[SSTR] + (n - k)\sigma^2 \\ &= (k - 1)\sigma^2 + \sum_{j=1}^k n_j(\mu_j - \mu)^2 + (n - k)\sigma^2 \end{aligned}$$

Exercise 2

```
setwd("C:/Users/merri/Documents/MATH-31H/MATH 181B/Homework 6")
teen = read.csv("Adolescent.csv")$x
adult = read.csv("Adult.csv")$x
middle = read.csv("MiddleAge.csv")$x

People = data.frame(times = c(teen, adult, middle),
                     Age= c(rep('Teen', length(teen)), rep('Adult', length(adult)),
                           rep('Middle', length(middle))))
summary(aov(times~Age, People))

# > summary(aov(times~Age, People))
#               Df Sum Sq Mean Sq F value    Pr(>F)
# Age           '2   173.6    86.80   22.46 3.08e-09 ***
# Residuals    147   568.1     3.86
# —————

# H0: mu_teen = mu_adult = mu_middle, H1: At least one mu is different
# Assume that Yij are independent, normal, and
# have the same variation for each j
# Since the p-value of 3.08e-09 is <0.05, we reject the null hypothesis and
# the mean screen times of all the age groups are not the same.

No, you cannot conclude that average adolescents spend more time in front
of the screen than the other two groups because the conclusion only says that at
least one group has a different mean. For the same reason, you cannot conclude
that adults spend more time than middle aged adults.

t.test(teen, adult, var.equal = T) # p-value = 1.667e-05
t.test(teen, middle, var.equal = T) # p-value = 2.448e-09
t.test(adult, middle, var.equal = T) # p-value = 0.04485

# Using the Bonferroni correction we use alpha/n = 0.01667
# Since mean of teen is greater than adult and middle and
# the p-value is <alpha/n, we conclude that teens have more
# screen time on average than adults and middle.
# Since 0.04485 > alpha/n, we cannot conclude that adults
# watch more tv than middle.
```

Exercise 3

```
reject <- function(p) {
  first = rnorm(50, 1, 1);
  second = rnorm(50, 1, 1);
  third = rnorm(50, 1, 1);
  p1 = t.test(first, second, var.equal = T)$p.value;
  p2 = t.test(second, third, var.equal = T)$p.value;
  p3 = t.test(first, third, var.equal = T)$p.value;
  return (p1<p | p2<p | p3<p);
}

regular_errors = rep(0, 1000);
bonferroni_errors = rep(0, 1000);
for(i in 1:1000) {
  regular_errors[i] = reject(0.05)
  bonferroni_errors[i] = reject(0.05/3)
}
sum(regular_errors)/1000
sum(bonferroni_errors)/1000
# > sum(regular_errors)/1000
# [1] 0.118
#
# > sum(bonferroni_errors)/1000
# [1] 0.042

# The proportion of type 1 errors is indeed <0.05 for the bonferroni correction,
# but it is more than double 0.05 without the correction.
```