Daily Check 8

2025-04-23

R. Markdown

Question: Daily Check 8

Answer this question: After running an ANOVA and obtaining a statistically significant p-value, we might want to know exactly WHICH means are different from each other. As discussed in class and in our textbook, why would it be a bad idea to simply do t-tests on each pair of means?

Also complete the two Your Turns, put all of this into an R Markdown file and submit both your pdf output and Rmd files.

Answer this question: After running an ANOVA and obtaining a statistically significant p-value, we might want to know exactly WHICH means are different from each other. As discussed in class and in our textbook, why would it be a bad idea to simply do t-tests on each pair of means?

It is a bad idea to simply do t-tests on each pair of means after a significant ANOVA because doing so does not control the experiment wise error rate. When you perform multiple t-tests, the probability of making at least one Type I error increases with the number of comparisons you make. This means you are more likely to incorrectly conclude that a difference exists when it actually does not. Instead, you should use special multiple comparison procedures that are designed to control the overall Type I error rate when making several pairwise comparisons.

```
google <- c(46, 49, 51, 42)
waze <- c(44, 47, 47, 43)
gut <- c(50, 51, 45, 43)
nav_df <- data.frame(
time = c(google, waze, gut),
method = c(rep("google", length(google)),
rep("waze", length(waze)),
rep("gut", length(gut)))
)</pre>
```

```
pan <- c(12.70, 12.30, 12.40, 12.80, 12.60,
12.50, 12.20, 12.70, 12.40, 12.60)
grilled <- c(11.80, 11.55, 11.75, 11.65, 11.50,
11.75, 11.30, 12.10, 11.50, 11.45)
broiled <- c(11.75, 11.85, 11.60, 11.80, 11.75,
11.65, 11.75, 11.70, 11.85, 11.75)
burgers_df <- data.frame(
sat_fat = c(pan, grilled, broiled),
method = c(rep("pan", length(pan)),
rep("grilled", length(grilled)),
rep("broiled", length(broiled)))
)</pre>
```

```
sample(nav_df$time)
  [1] 43 45 50 44 51 46 47 51 42 43 49 47
set.seed(10242024)
nav_perm <- nav_df # temporarily copy the whole df</pre>
nav_perm$time <- sample(nav_perm$time)</pre>
perm_anova <- aov(time ~ method, data=nav_perm)</pre>
summary(perm_anova)
##
               Df Sum Sq Mean Sq F value Pr(>F)
## method
               2 45.5
                           22.75 3.033 0.0984 .
## Residuals
              9
                    67.5
                            7.50
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
names(perm_anova)
## [1] "coefficients" "residuals"
                                         "effects"
## [5] "fitted.values" "assign"
                                         "qr"
                                                         "df.residual"
## [9] "contrasts"
                        "xlevels"
                                         "call"
                                                         "terms"
## [13] "model"
summary(perm_anova)[[1]][1,4]
## [1] 3.033333
# Create a vector to store the F values
perm_f <- NA
# set the number of replications
reps <- 10000
# do it lots of times in a loop
for(i in 1:reps){
nav_perm <- nav_df</pre>
nav_perm$time <- sample(nav_perm$time)</pre>
perm_f[i] <- summary(aov(time ~ method, data=nav_perm))[[1]][1,4]</pre>
}
summary(aov(time ~ method, data=nav_df))[[1]][1,4]
## [1] 0.4130435
F <- summary(aov(time ~ method, data=nav_df))[[1]][1,4]
sum(perm_f >= F)/reps
## [1] 0.6527
```

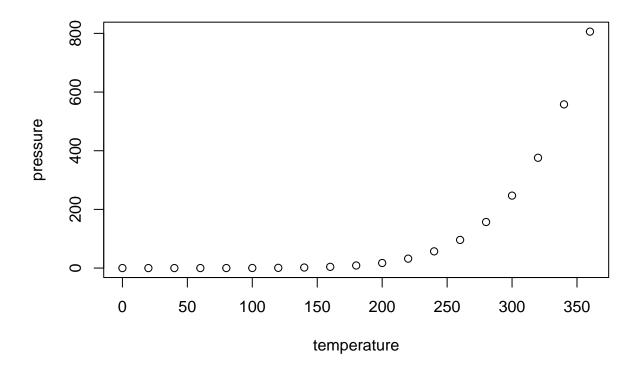
Your Turn #1

```
set.seed(4232025)
data(PlantGrowth)
#Create a vector to store the F values
perm_f <- NA
# set reps
reps <- 10000
#do it lots of times
for(i in 1: reps)
{
  plant_perm <- PlantGrowth</pre>
  plant_perm$weight <- sample(plant_perm$weight)</pre>
  perm_f[i] <- summary(aov(weight ~ group, data=plant_perm))[[1]][1,4]</pre>
}
F <- summary(aov(weight ~ group, data=PlantGrowth))[[1]][1,4]
sum(perm_f >= F)/reps
## [1] 0.0163
Here is the ANOVA analysis:
model1 <- aov(weight ~ group, data = PlantGrowth)</pre>
summary(model1)
               Df Sum Sq Mean Sq F value Pr(>F)
##
## group
               2 3.766 1.8832 4.846 0.0159 *
## Residuals 27 10.492 0.3886
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
library(dplyr)
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
summary <- burgers_df %>% group_by(method) %>%
summarize(means=mean(sat_fat))
print(summary)
## # A tibble: 3 x 2
     method means
```

```
## <chr> <dbl>
## 1 broiled 11.7
## 2 grilled 11.6
## 3 pan
              12.5
summary(aov(sat_fat ~ method, data=burgers_df))
##
              Df Sum Sq Mean Sq F value
                                           Pr(>F)
## method
               2 4.653 2.3266
                                    73.6 1.17e-11 ***
## Residuals
              27 0.853 0.0316
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
q <- abs(summary$means[1] - summary$means[2]) / sqrt(summary(aov(sat_fat~method, data = burgers_df))[[1]
print(q)
## [1] 1.956468
ptukey(q, nmeans=3, df=27, lower.tail=FALSE)
## [1] 0.3634447
\#\# Your Turn 2
library(dplyr)
summary <- PlantGrowth %>% group_by(group) %>% summarize(means = mean(weight))
print(summary)
## # A tibble: 3 x 2
##
     group means
##
    <fct> <dbl>
## 1 ctrl 5.03
## 2 trt1 4.66
## 3 trt2
          5.53
q <- abs(summary$means[2] - summary$means[3]) / sqrt(summary(aov(weight ~ group, data = PlantGrowth))[[
ptukey(q, nmeans = 3, df = 27, lower.tail = FALSE)
## [1] 0.01200642
All pairwise p-values:
TukeyHSD(model1)
##
     Tukey multiple comparisons of means
##
       95% family-wise confidence level
##
```

Including Plots

You can also embed plots, for example:



Note that the echo = FALSE parameter was added to the code chunk to prevent printing of the R code that generated the plot.