# STAT 8010 R Session 4

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# Contents

Session Objectives	1
Two sample t-test	2
Tapeworm example	2
Two sample t test with only sample statistics	4
Paired t-Test	Ę
AONVA	7
Toy Examples	7
F Distribution	ç
Effects of Ethanol on Sleep Time Example	10
Facebook Example	12
Multiple Comparisons	14
Fisher's LSD	14
Tukey's HSD	15
CRD and RCBD	15
Create the data set	15
Two-way ANOVA	15
One-way ANOVA	16
Interaction plot: assessing the additivity assumption	16

# Session Objectives

- To gain experience with R, a programming language and free software environment for statistical computing and graphics.
- $\bullet\,$  To perform two sample t-test and paired t-test using R
- To conduct ANOVA and mulitple comparisons using R

# Two sample t-test

### Tapeworm example

```
treatment <- c(18, 43, 28, 50, 16, 32, 13, 35, 38, 33, 6, 7)
control <- c(40, 54, 26, 63, 21, 37, 39, 23, 48, 58, 28, 39)
dat <- data.frame(cbind(treatment, control))</pre>
summary(dat)
##
      treatment
                       control
##
   Min. : 6.00
                   Min.
                           :21.00
##
   1st Qu.:15.25
                   1st Qu.:27.50
## Median :30.00
                  Median :39.00
          :26.58
                   Mean :39.67
## Mean
## 3rd Qu.:35.75
                   3rd Qu.:49.50
           :50.00
                           :63.00
## Max.
                   Max.
boxplot(dat, boxwex = 0.3, las = 1)
60
50
40
30
20
10
                  treatment
                                                  control
apply(dat, 2, mean)
## treatment
               control
## 26.58333 39.66667
apply(dat, 2, sd)
## treatment
               control
## 14.36193 13.85859
var.test(treatment, control)
##
```

## F test to compare two variances

```
##
## data: treatment and control
## F = 1.074, num df = 11, denom df = 11, p-value = 0.9079
## alternative hypothesis: true ratio of variances is not equal to 1
## 95 percent confidence interval:
## 0.3091686 3.7306092
## sample estimates:
## ratio of variances
##
             1.073959
# Assuming $\sigma_{1}=\sigma_{2}$
t.test(treatment, control, var.equal = T)
##
  Two Sample t-test
##
##
## data: treatment and control
## t = -2.2709, df = 22, p-value = 0.03329
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -25.031761 -1.134906
## sample estimates:
## mean of x mean of y
## 26.58333 39.66667
# Assuming $\sigma_{1}\neq sigma_{2}$
t.test(treatment, control, var.equal = F)
##
## Welch Two Sample t-test
## data: treatment and control
## t = -2.2709, df = 21.972, p-value = 0.03331
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -25.032642 -1.134025
## sample estimates:
## mean of x mean of y
## 26.58333 39.66667
# Left-tailed test
t.test(treatment, control, alternative = "less")
##
## Welch Two Sample t-test
## data: treatment and control
## t = -2.2709, df = 21.972, p-value = 0.01665
## alternative hypothesis: true difference in means is less than 0
## 95 percent confidence interval:
##
         -Inf -3.189613
## sample estimates:
## mean of x mean of y
## 26.58333 39.66667
```

#### Two sample t test with only sample statistics

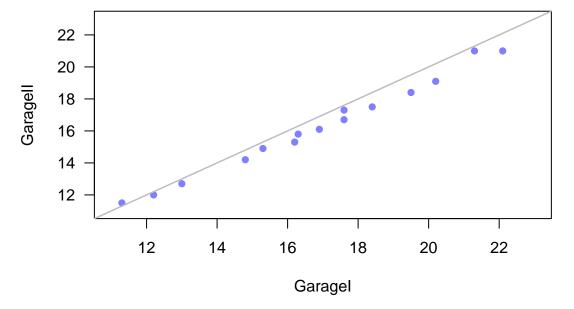
```
t.test.from.summary.data <- function(mean1, sd1, n1, mean2, sd2, n2, ...) {
    data1 \leftarrow scale(1:n1)*sd1 + mean1
    data2 <- scale(1:n2)*sd2 + mean2</pre>
    t.test(data1, data2, ...)
}
t.test.from.summary.data(12.5, 7.63, 10, 27.5, 15.3, 10)
##
##
    Welch Two Sample t-test
##
## data: data1 and data2
## t = -2.7744, df = 13.216, p-value = 0.01558
\mbox{\tt \#\#} alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -26.660768 -3.339232
## sample estimates:
## mean of x mean of y
##
        12.5
                   27.5
t.test.from.summary.data(19.45, 4.3, 37, 18.2, 2.2, 31)
##
   Welch Two Sample t-test
##
## data: data1 and data2
## t = 1.5435, df = 55.507, p-value = 0.1284
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.3726447 2.8726447
## sample estimates:
## mean of x mean of y
       19.45
                 18.20
## Check
(\mathbf{df} = ((4.3^2)/37 + (2.2^2)/31)^2 / (((4.3^2)/37)^2 / 36 + ((2.2^2)/31)^2 / 30))
## [1] 55.50703
(se \leftarrow sqrt(4.3^2 / 37 + 2.2^2 / 31))
## [1] 0.8098511
(tstat \leftarrow (19.45 - 18.2) / se)
## [1] 1.543494
```

```
(Pvalue <- 2 * (1 - pt(1.5435, df)))
```

## [1] 0.128392

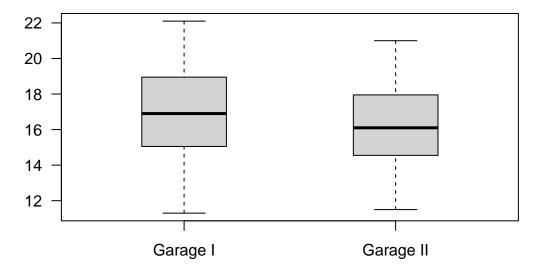
# Paired t-Test

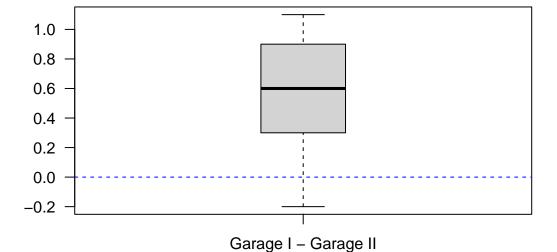
```
## GarageI GarageII
## 16.84667 16.23333
```



```
t.test(GarageI, GarageII, alternative = c("greater"), var.equal = F)
```

```
##
## Welch Two Sample t-test
##
```



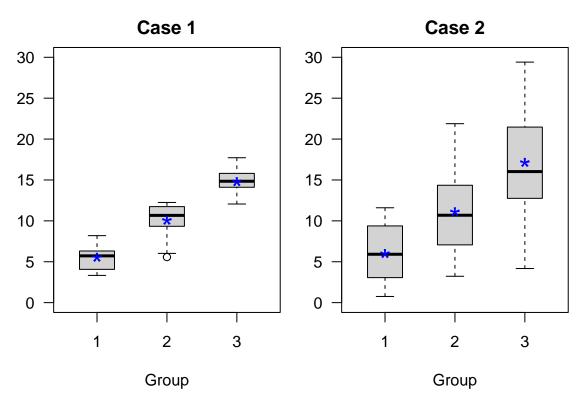


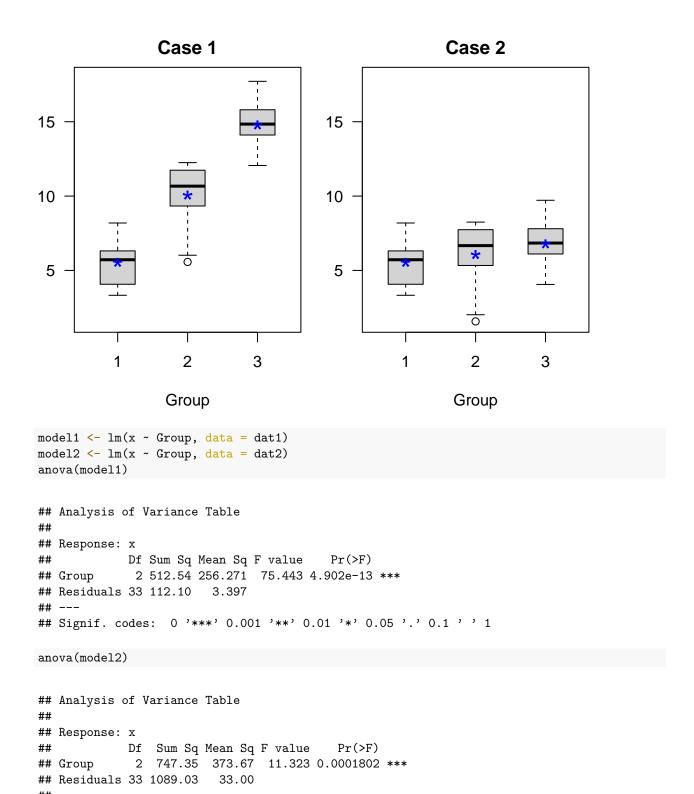
```
t.test(GarageI, GarageII, alternative = c("greater"), paired = T)
```

### **AONVA**

#### Toy Examples

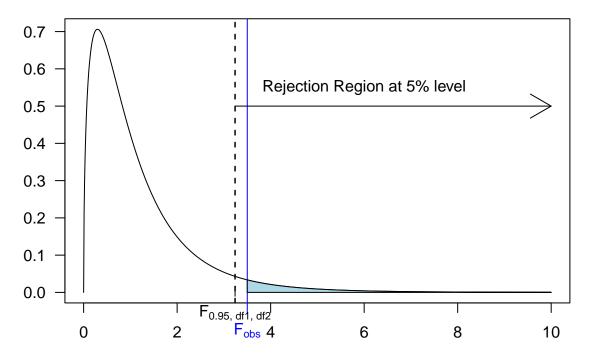
```
set.seed(1)
base1 \leftarrow rnorm(n = 36, sd = 2)
base2 \leftarrow rnorm(n = 36, sd = 6)
dat1 \leftarrow base1 + c(rep(5, 12), rep(10, 12), rep(15, 12))
dat2 \leftarrow base2 + c(rep(5, 12), rep(10, 12), rep(15, 12))
dat3 \leftarrow base1 + rep(5:7, each = 12)
level <- as.factor(rep(1:3, each = 12))</pre>
dat1 <- data.frame(x = dat1, Group = level)</pre>
dat2 <- data.frame(x = dat2, Group = level)</pre>
dat3 <- data.frame(x = dat3, Group = level)</pre>
library(dplyr)
g1summary <- dat1 %>%
select(x, Group) %>%
group_by(Group) %>%
summarise(mean = mean(x), sd1 = sd(x))
g2summary <- dat2 %>%
select(x, Group) %>%
group_by(Group) %>%
summarise(mean = mean(x), sd1 = sd(x))
g3summary <- dat3 %>%
select(x, Group) %>%
group_by(Group) %>%
summarise(mean = mean(x), sd1 = sd(x))
par(mfrow = c(1, 2), mar = c(4.1, 2.1, 2.1, 1.1))
boxplot(x ~ Group, data = dat1, las = 1, boxwex = 0.5,
        ylab = "", ylim = c(0, 30), main = "Case 1")
for (i in 1:3) points(i, g1summary$mean[i], pch = "*",
                       col = "blue", cex = 2)
boxplot(x ~ Group, data = dat2, las = 1, boxwex = 0.5,
```





## F Distribution

## Signif. codes: 0 '\*\*\* 0.001 '\*\* 0.01 '\* 0.05 '.' 0.1 ' 1



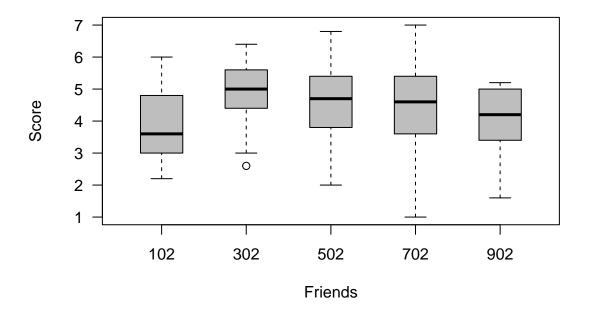
Effects of Ethanol on Sleep Time Example

```
### Data setup
set.seed(124)
g1 <- rnorm(5, 83, 9); g2 <- rnorm(5, 76, 9.5); g3 <- rnorm(5, 73, 9.2); g4 <- rnorm(5, 70, 9)
dat <- cbind(Response = c(g1, g2, g3, g4), Treatment = as.factor(rep(1:4, each = 5)))
dat <- data.frame(dat)
dat$Treatment <- as.factor(dat$Treatment)
par(mar = c(4.1, 4.1, 1.1, 1.1))
boxplot(Response ~ Treatment, data = dat, horizontal = T, yaxt = "n", ylab = "", xlab = "")
axis(2, at = 1:4, labels = c("Control", "1g/kg", "2g/kg", "4g/kg"), las = 1)</pre>
```

```
### Data Summary
summary <- dat %>%
select(Response, Treatment) %>%
group_by(Treatment) %>%
summarise(mean = mean(Response),
          sd1 = sd(Response))
lm <- lm(Response ~ Treatment, dat)</pre>
anova(lm)
## Analysis of Variance Table
##
## Response: Response
            Df Sum Sq Mean Sq F value Pr(>F)
## Treatment 3 861.13 287.044 4.2542 0.02173 *
## Residuals 16 1079.56 67.472
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
### Pairwise t-test
t.test(dat$Response[1:5], dat$Response[6:10], var.equal = T)
```

```
##
## Two Sample t-test
##
## data: dat$Response[1:5] and dat$Response[6:10]
## t = 0.24012, df = 8, p-value = 0.8163
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -10.09426 12.44081
## sample estimates:
## mean of x mean of y
## 82.15052 80.97724
```

```
t.test(dat$Response[1:5], dat$Response[6:10], var.equal = F)
##
## Welch Two Sample t-test
## data: dat$Response[1:5] and dat$Response[6:10]
## t = 0.24012, df = 6.2015, p-value = 0.818
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -10.68922 13.03577
## sample estimates:
## mean of x mean of y
## 82.15052 80.97724
Facebook Example
dat <- read.csv("FacebookFriends.csv")</pre>
head(dat); str(dat)
##
    Friends Participant Score
## 1
       102
                     1
                         3.8
## 2
                     2 3.6
        102
## 3
       102
                     3 3.2
## 4
                     4 2.4
        102
## 5
        102
                     5 4.8
## 6
        102
                       3.0
## 'data.frame': 134 obs. of 3 variables:
## $ Friends
               ## $ Participant: int 1 2 3 4 5 6 7 8 9 10 ...
             : num 3.8 3.6 3.2 2.4 4.8 3 4.2 3.6 3.2 3 ...
dat$Friends <- as.factor(dat$Friends)</pre>
str(dat)
## 'data.frame':
                  134 obs. of 3 variables:
## $ Friends : Factor w/ 5 levels "102", "302", "502", ...: 1 1 1 1 1 1 1 1 1 1 ...
## $ Participant: int 1 2 3 4 5 6 7 8 9 10 ...
## $ Score : num 3.8 3.6 3.2 2.4 4.8 3 4.2 3.6 3.2 3 ...
boxplot(Score ~ Friends, data = dat, las = 1, col = "gray", boxwex = 0.5)
```



```
library(dplyr)
summary <- dat %>%
select(Score, Friends) %>%
group_by(Friends) %>%
summarise(mean = mean(Score),
          sd1 = sd(Score))
summary
## # A tibble: 5 x 3
    Friends mean sd1
##
     <fct> <dbl> <dbl>
##
              3.82 0.999
## 1 102
## 2 302
              4.88 0.851
## 3 502
              4.56 1.07
## 4 702
              4.41 1.43
## 5 902
              3.99 1.02
lm <- lm(Score ~ Friends, dat)</pre>
anova(lm)
## Analysis of Variance Table
##
## Response: Score
              Df Sum Sq Mean Sq F value Pr(>F)
```

4.142 0.00344 \*\*

## Signif. codes: 0 '\*\*\* 0.001 '\*\* 0.01 '\* 0.05 '.' 0.1 ' ' 1

4 19.89 4.9726

## Residuals 129 154.87 1.2005

## Friends

par(mfrow = c(1, 3))
plot(lm, which = 1:3)

## ---

```
Residuals vs Fitted
                                                                   Normal Q-Q
                                                                                                                    Scale-Location
                                                          က
                                                                                                                                   0102
087
                                                                                                                                   0960
                             8
                                                                                                                                   0000
                                000
                                                                                                      /IStandardized residuals
                             0000
                                                   Standardized residuals
                                                                                                                                   0
Residuals
                             0
                                                                                                                                   0
                                                                                                                                       8
                                                          0
                  0000
                                                                                                                                       8
                                                                                                                                   0
                                000
                             00000
                                                                                                                                   0
                                8
                                                          ī
                  0
                                                                                                                                   0
       7
                                                          7
                                 0
       က
                             O<sub>87</sub> O<sub>102</sub>
                                                                                                              0.0
       4
                                 4.6
                                                                               0
                                                                                         2
            3.8
                      4.2
                                                                   -2
                                                                                    1
                                                                                                                   3.8
                                                                                                                             4.2
                                                                                                                                        4.6
                    Fitted values
                                                                  Theoretical Quantiles
                                                                                                                           Fitted values
```

```
aov <- aov(Score ~ Friends, dat)
aov</pre>
```

```
## Call:
## aov(formula = Score ~ Friends, data = dat)
##
## Terms:
## Friends Residuals
## Sum of Squares 19.89023 154.86679
## Deg. of Freedom 4 129
##
## Residual standard error: 1.095681
## Estimated effects may be unbalanced
```

# Multiple Comparisons

## Fisher's LSD

```
library(agricolae)
LSD_none <- LSD.test(aov ,"Friends", p.adj = "none")
LSD_none$groups</pre>
```

```
## Score groups
## 302 4.878788 a
## 502 4.561538 ab
## 702 4.406667 abc
## 902 3.990476 bc
## 102 3.816667 c
```

```
LSD_bon <- LSD.test(aov ,"Friends", p.adj = "bonferroni")
LSD_bon$groups
```

```
## Score groups
## 302 4.878788 a
## 502 4.561538 ab
## 702 4.406667 ab
## 902 3.990476 b
## 102 3.816667 b
```

### Tukey's HSD

```
HSD <- TukeyHSD(aov, conf.level = 0.95)
HSD$Friends</pre>
```

```
## diff lwr upr p adj
## 302-102 1.0621212 0.2488644 1.87537798 0.003889635
## 502-102 0.7448718 -0.1132433 1.60298691 0.121456224
## 702-102 0.5900000 -0.2402014 1.42020143 0.288431585
## 902-102 0.1738095 -0.7320145 1.07963355 0.984016816
## 502-302 -0.3172494 -1.1121910 0.47769215 0.804080046
## 702-302 -0.4721212 -1.2368466 0.29260420 0.432633745
## 902-302 -0.8883117 -1.7345313 -0.04209203 0.034535577
## 702-502 -0.1548718 -0.9671402 0.65739661 0.984391504
## 902-702 -0.4161905 -1.2787075 0.44632652 0.669927748
```

#### CRD and RCBD

Create the data set

```
x <- c(52, 47, 44, 51, 42, 60, 55, 49, 52, 43, 56, 48, 45, 44, 38)
trt <- rep(c("A", "B", "C"), each = 5)
blk <- rep(1:5, 3)
dat <- data.frame(x = x, trt = trt, blk = as.factor(blk))</pre>
```

#### Two-way ANOVA

```
aov <- aov(x ~ trt + blk, data = dat)
lm <- lm(x ~ trt + blk, data = dat)
anova(lm)

## Analysis of Variance Table
##
## Response: x
## Df Sum Sq Mean Sq F value Pr(>F)
```

```
## trt     2   89.2   44.60   7.6239   0.0140226 *
## blk     4   363.6   90.90   15.5385   0.0007684 ***
## Residuals     8   46.8     5.85
## ---
## Signif. codes:     0 '***'   0.001 '**'   0.05 '.'   0.1 ' ' 1
```

### One-way ANOVA

```
lm2 <- lm(x ~ trt, data = dat)
anova(lm2)

## Analysis of Variance Table

##
## Response: x
## Df Sum Sq Mean Sq F value Pr(>F)
## trt 2 89.2 44.6 1.3041 0.3073
## Residuals 12 410.4 34.2
```

# Interaction plot: assessing the additivity assumption

```
interaction.plot(dat$trt, dat$blk, x, las = 1, col = 1:5)
```

