# Statistics Examen NeuroBIM

Maxime Houtekamer
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# Loading the data file into R

First of all, the data was loaded into R from the textfile. A summary of the data was obtained as an indication whether the data was correctly loaded into R.

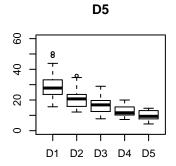
### Let's add a collumn for the difference between time needed

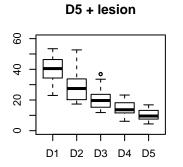
# Creating seperate files for the 4 conditions

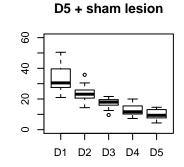
The mice were either trained in 3 sessions (D3) or in 5 sessions (D5). Within each of these two groups, the animals were either lesioned in the dorsal hippocampus (H) or they were given a sham lesion (SH). These groups were originally stored in the datafile, but will now be sorted in order to easily be able to display them seperately.

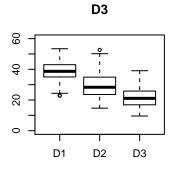
### Learning time

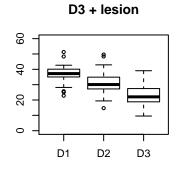
The mice were given a task, and the time they spent in the dark is a measure of how well they learnt it.

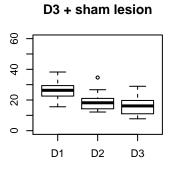












- ## [1] 29.5587
- ## [1] 22.88382
- ## [1] 13.70844
- ## [1] 10.73177

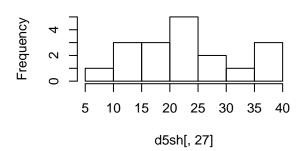
# Is the data normally distributed?

First, Let's look at histograms of each group.

# Histogram of d5h[, 27]

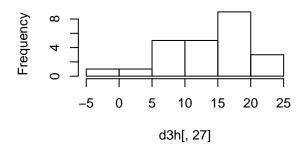
# Preduction of the state of the

# Histogram of d5sh[, 27]

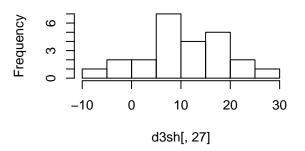


### Histogram of d3h[, 27]

d5h[, 27]



### Histogram of d3sh[, 27]



### NORMAL DISTIBUTION?

We will carry out the shapiro-Wilk test. If p>a (bigger than 0.05 generally), the data is normal.

```
##
##
    Shapiro-Wilk normality test
##
## data: d5h[, 27]
## W = 0.96506, p-value = 0.5482
##
    Shapiro-Wilk normality test
##
##
## data: d5sh[, 27]
## W = 0.94456, p-value = 0.3461
##
    Shapiro-Wilk normality test
##
## data: d3h[, 27]
## W = 0.95398, p-value = 0.3297
##
```

```
## Shapiro-Wilk normality test
##
## data: d3sh[, 27]
## W = 0.98565, p-value = 0.9735
```

All the values are higher than p=0.05, so the data is normally distributed.

# Making a new dataframe for ANOVA

We will make a list of the factors (d5h, d5sh, d3h, d3sh), and a list with the "learned" decrease in time needed to explore the matrix.

### Homogeneity of Variance

```
##
## Bartlett test of homogeneity of variances
##
## data: d1$values by d1$exp
## Bartlett's K-squared = 3.6572, df = 3, p-value = 0.3009
```

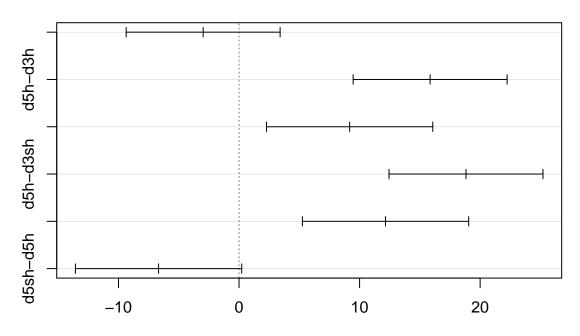
### ANOVA

Maybe we should instead to a repeated measures anova where we follow the animal over the different learning trials.

```
## Df Sum Sq Mean Sq F value Pr(>F)
## d1\( \)exp 3 5260 1753.3 24.6 1.38e-11 ***
## Residuals 86 6129 71.3
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
```

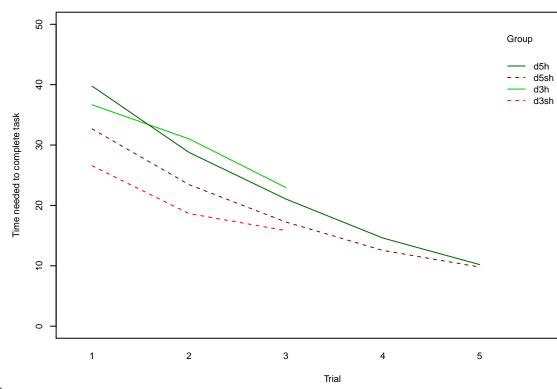
The anova is highly significant at p<0.05. Let's do a post-hoc Tukey test to find where the differences are #Tukey posthoc

```
##
     Tukey multiple comparisons of means
##
       95% family-wise confidence level
##
## Fit: aov(formula = f1)
##
## $ d1$exp
##
                  diff
                              lwr
                                         upr
## d3sh-d3h
            -2.976667
                        -9.361469 3.4081355 0.6150204
## d5h-d3h
             15.850260
                         9.465458 22.2350626 0.0000000
                         2.279007 16.0717565 0.0042380
## d5sh-d3h
              9.175382
## d5h-d3sh 18.826927
                        12.442125 25.2117293 0.0000000
## d5sh-d3sh 12.152049
                         5.255674 19.0484232 0.0000790
## d5sh-d5h -6.674878 -13.571253 0.2214961 0.0614873
```



Differences in mean levels of d1\$exp

### Interaction plot of learning



### #Repeated measures ANOVA

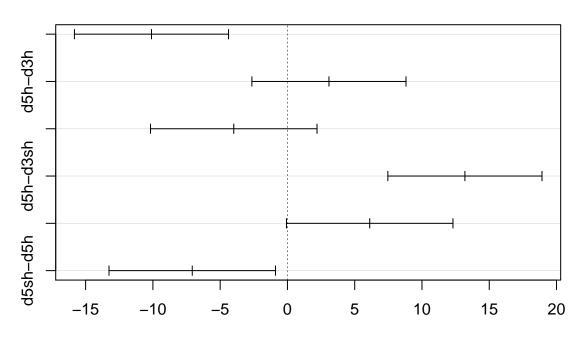
```
##
## Error: id
            Df Sum Sq Mean Sq F value
##
                                         Pr(>F)
## group
                  5594 1864.7
                                 27.11 1.95e-12 ***
## Residuals 86
                  5914
                          68.8
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Error: Within
##
              Df Sum Sq Mean Sq F value Pr(>F)
                  22320
                            5580 169.998 < 2e-16 ***
## time
                     751
                              94
                                   2.861 0.00462 **
## group:time
                8
                    8272
## Residuals 252
                              33
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Loading required package: lme4
## Loading required package: Matrix
## Loading required package: reshape2
## Loading required package: lsmeans
## Loading required package: estimability
## Attaching package: 'lsmeans'
## The following object is masked from 'package:base':
```

```
##
##
       rbind
##
## *******
## Welcome to afex. Important changes in the current version:
## - Functions for ANOVAs have been renamed to: aov car(), aov ez(), and aov 4().
## - ANOVA functions return an object of class 'afex aov' as default, see: ?aov car
## - 'afex_aov' objects can be passed to 1smeans for contrasts and follow-up tests.
## - Reset previous (faster) behavior via: afex_options(return_aov='nice')
## - Many more arguments can now be set globally via options, see: afex_options()
## *******
## Loading required package: mvtnorm
## Loading required package: survival
## Loading required package: TH.data
model.cs <- gls(tasktime \sim group * time , data = d2, corr = corCompSymm(, form = \sim 1 \mid id)
summary(model.cs)
```

The between group tests indicates that the variable group is significant. consequently, int he graph we see that the liens for the two groups are rather far apart. The within subject test indicates that there is a significant time effect, inother words, the groups do change other time, both groups are taking less time to complete the task over time. Morover, the interaction of time and group is significant which means that the groups are changing over time but are changing in different ways, which means that in the graph, the liens will not be parallel.

### DAY 1

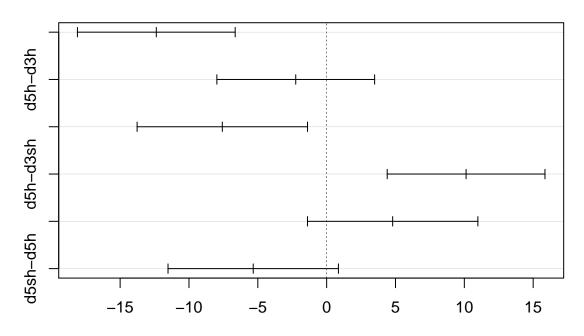
```
##
               Df Sum Sq Mean Sq F value
                                           Pr(>F)
## D1$exp
                    2324
                           774.7
                                    13.5 2.68e-07 ***
                3
## Residuals
               86
                    4934
                            57.4
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
     Tukey multiple comparisons of means
       95% family-wise confidence level
##
##
## Fit: aov(formula = fD1)
##
## $`D1$exp`
##
                   diff
                                 lwr
                                            upr
                                                    p adj
## d3sh-d3h
            -10.104688 -15.83327806 -4.3760969 0.0000775
## d5h-d3h
               3.086062 -2.64252806 8.8146531 0.4957580
## d5sh-d3h
              -3.989444 -10.17702945 2.1981406 0.3356221
## d5h-d3sh
              13.190750
                          7.46215944 18.9193406 0.0000002
               6.115243 -0.07234195 12.3028281 0.0539491
## d5sh-d3sh
## d5sh-d5h
              -7.075507 -13.26309195 -0.8879219 0.0184144
```



Differences in mean levels of D1\$exp

### DAY 2

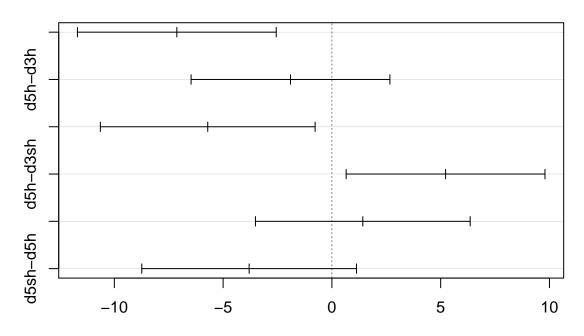
```
Df Sum Sq Mean Sq F value
                                           Pr(>F)
## D2$exp
                3
                    2191
                           730.4
                                   12.73 5.84e-07 ***
## Residuals
              86
                    4935
                            57.4
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
     Tukey multiple comparisons of means
##
##
       95% family-wise confidence level
##
## Fit: aov(formula = fD2)
##
## $`D2$exp`
##
                   diff
                               lwr
                                          upr
            -12.373333 -18.102674 -6.6439923 0.0000012
## d3sh-d3h
## d5h-d3h
              -2.244844 -7.974185 3.4844973 0.7343947
              -7.578073 -13.766469 -1.3896773 0.0099558
## d5sh-d3h
## d5h-d3sh
              10.128490
                         4.399149 15.8578307 0.0000745
## d5sh-d3sh
              4.795260 -1.393135 10.9836561 0.1850429
## d5sh-d5h
              -5.333229 -11.521625 0.8551665 0.1161882
```



Differences in mean levels of D2\$exp

```
#DAY 3
```

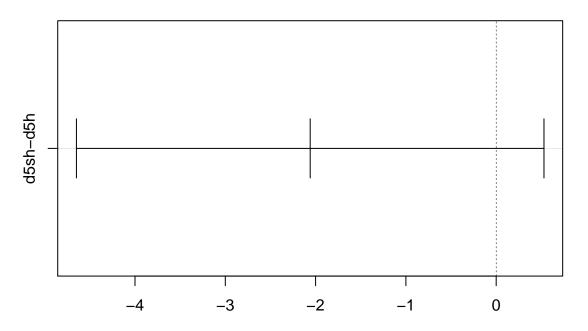
```
Df Sum Sq Mean Sq F value
## D3$exp
               3 758.6
                          252.8
                                  6.927 0.000313 ***
## Residuals
              86 3139.3
                           36.5
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
     Tukey multiple comparisons of means
##
##
      95% family-wise confidence level
##
## Fit: aov(formula = fD3)
##
## $\D3\exp\
##
                 diff
                             lwr
                                        upr
## d3sh-d3h
            -7.128021 -11.697564 -2.5584780 0.0005588
## d5h-d3h
            -1.903073
                      -6.472616 2.6664700 0.6958769
## d5sh-d3h -5.706979 -10.642650 -0.7713088 0.0167529
## d5h-d3sh
             5.224948
                        0.655405 9.7944908 0.0184241
## d5sh-d3sh 1.421042
                       -3.514629 6.3567121 0.8745762
## d5sh-d5h -3.803906 -8.739577 1.1317642 0.1889809
```



Differences in mean levels of D3\$exp

### $\#\mathrm{DAY}\ 4$

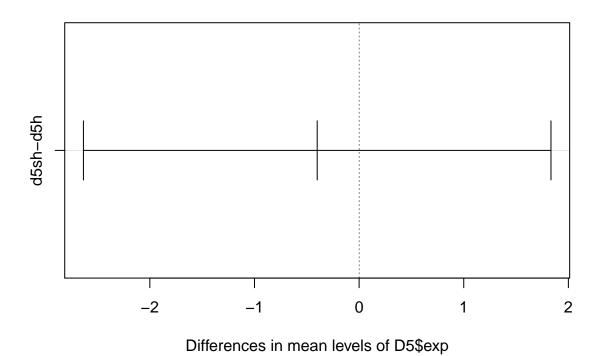
```
Df Sum Sq Mean Sq F value Pr(>F)
## D4$exp
                1
                    43.6
                           43.64
                                   2.586 0.116
               40 675.1
                           16.88
## Residuals
##
     Tukey multiple comparisons of means
       95% family-wise confidence level
##
##
## Fit: aov(formula = fD4)
##
## $`D4$exp`
##
                 diff
                            lwr
## d5sh-d5h -2.059722 -4.648644 0.5291991 0.1157122
```



Differences in mean levels of D4\$exp

```
\#\mathrm{DAY}\ 5
```

```
Df Sum Sq Mean Sq F value Pr(>F)
## D5$exp
               1
                     1.7
                          1.651
                                0.131 0.719
              40 502.8 12.571
## Residuals
##
     Tukey multiple comparisons of means
       95% family-wise confidence level
##
##
## Fit: aov(formula = fD5)
##
## $`D5$exp`
##
                 diff
                            lwr
## d5sh-d5h -0.4006285 -2.634966 1.833709 0.7189687
```



Are there structures that are differentially activated depending on the duration of the training?

I think the best option here is MANOVA. I used this video first https://www.youtube.com/watch?v=48cZ2cMBpio

```
## Df Pillai approx F num Df den Df Pr(>F)
## as.factor(TRAIN) 1 0.87467 8.0807 19 22 4.842e-06 ***
## Residuals 40
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
```

I don't think these results are helpful at all. So there are areas that differ.. ok..

# What are the structure activities that are correlated with performance (in the last training session)?

Ok. So we need a correlation. then we need a matrix.

```
## Loading required package: grid
## Loading required package: Formula
## Loading required package: ggplot2
##
```

```
## Attaching package: 'Hmisc'
##
  The following objects are masked from 'package:base':
##
       format.pval, round.POSIXt, trunc.POSIXt, units
##
              row column
                                 cor
## 172
             STLD final -0.25688878 0.01451892
             STMD final -0.11459476 0.28215434
## 173
## 174
         AMBASLAT final 0.12201443 0.25194559
                  final 0.03474115 0.74512462
## 175
           AMLAT
## 176
          ENTORH final -0.08409247 0.43068690
## 177
           PERIRH final -0.06621543 0.53520851
## 178
              CA1 final -0.29772902 0.05549844
## 179
              CA3 final -0.30402470 0.05029389
## 180
               DG final -0.15854518 0.31593655
## 181
         CINGULAR
                  final -0.23694505 0.02454052
         PRELIMB
## 182
                  final 0.06786582 0.52505980
## 183
          SOMSENS
                  final
                          0.09701581 0.36300415
## 184
        SUBICULUM
                  final 0.24090412 0.02218022
## 185
         ACCCORE
                  final -0.08076198 0.44922795
## 186
         ACCSHELL
                   final -0.06811436 0.52353991
                   final 0.05980199 0.57554592
## 187
           VISUAL
## 188
         PIRIFORM final -0.05737391 0.59117707
         PARIETAL final 0.06761815 0.52657656
## 189
## 190 RETROSPLEN final 0.23501243 0.02576832
##
                       column
              row
                                       cor
## 121
             STLD data1[, 26]
                               0.420943916 0.04052046
## 122
             STMD data1[, 26]
                               0.508609712 0.01115149
## 123
         AMBASLAT data1[, 26]
                               0.314945042 0.13387549
## 124
           AMLAT data1[, 26]
                               0.153313786 0.47446618
           ENTORH data1[, 26]
## 125
                               0.022545485 0.91671974
          PERIRH data1[, 26] -0.029052418 0.89280380
## 126
## 127
         CINGULAR data1[, 26]
                               0.275005847 0.19340416
         PRELIMB data1[, 26] -0.273405820 0.19611358
## 128
          SOMSENS data1[, 26] -0.005872661 0.97827284
## 129
## 130
        SUBICULUM data1[, 26] -0.185807616 0.38469403
         ACCCORE data1[, 26]
## 131
                               0.351101816 0.09251817
## 132
         ACCSHELL data1[, 26]
                               0.088317879 0.68153189
           VISUAL data1[, 26]
## 133
                               0.062166829 0.77290674
## 134
         PIRIFORM data1[, 26]
                               0.175138682 0.41304156
## 135
         PARIETAL data1[, 26]
                               0.292301059 0.16573976
## 136 RETROSPLEN data1[, 26]
                               0.003533508 0.98692608
## Warning in sqrt(npair - 2): NaNs produced
##
                       column
              row
                                        cor
## 172
             STLD data1[, 24] -0.076227695 0.723325284
## 173
             STMD data1[, 24] -0.002770538 0.989748891
                              0.225778431 0.288775563
## 174
         AMBASLAT data1[, 24]
## 175
           AMLAT data1[, 24] 0.246296376 0.245979988
           ENTORH data1[, 24] -0.179720566 0.400729262
## 176
```

```
PERIRH data1[, 24] -0.532149076 0.007435526
## 177
## 178
              CA1 data1[, 24]
                                        NΑ
## 179
              CA3 data1[, 24]
                                        NA
## 180
              DG data1[, 24]
                                        NA
                                                    NΑ
## 181
         CINGULAR data1[, 24] -0.004120424 0.984754725
         PRELIMB data1[, 24] -0.236160472 0.266572650
## 182
          SOMSENS data1[, 24] 0.205784410 0.334698112
## 183
        SUBICULUM data1[, 24] -0.109996065 0.608889819
## 184
## 185
         ACCCORE data1[, 24]
                              0.075546660 0.725703792
## 186
         ACCSHELL data1[, 24]
                              0.177521572 0.406612533
## 187
          VISUAL data1[, 24] -0.021707390 0.919805660
         PIRIFORM data1[, 24]
                              0.150478333 0.482773527
## 188
## 189
        PARIETAL data1[, 24] 0.236988276 0.264850721
## 190 RETROSPLEN data1[, 24] -0.021632330 0.920082088
##
             row
                       column
## 172
             STLD data1[, 24] -0.32145479 0.12559258
## 173
             STMD data1[, 24] -0.28133541 0.18293688
## 174
         AMBASLAT data1[, 24] -0.25245550 0.23398786
           AMLAT data1[, 24] -0.21934773 0.30309215
## 175
          ENTORH data1[, 24] -0.44434154 0.02960353
## 176
           PERIRH data1[, 24] -0.30310613 0.14993311
## 177
## 178
              CA1 data1[, 24] -0.14480291 0.49961947
## 179
              CA3 data1[, 24] -0.10614695 0.62155328
              DG data1[, 24] 0.06378335 0.76715843
## 180
         CINGULAR data1[, 24] -0.45524520 0.02539334
## 181
## 182
         PRELIMB data1[, 24] 0.06031009 0.77952358
## 183
          SOMSENS data1[, 24] -0.22309427 0.29469885
        SUBICULUM data1[, 24] -0.36111617 0.08296873
## 184
          ACCCORE data1[, 24] -0.35409132 0.08958555
## 185
## 186
         ACCSHELL data1[, 24] -0.06335748 0.76867166
## 187
          VISUAL data1[, 24] 0.41607288 0.04314844
         PIRIFORM data1[, 24] -0.36375323 0.08058232
## 188
## 189
         PARIETAL data1[, 24] -0.29271892 0.16510766
## 190 RETROSPLEN data1[, 24] -0.17931114 0.40182103
##
                       column
             row
                                      cor
## 172
             STLD data1[, 26] 0.35961094 0.14272193
             STMD data1[, 26]
## 173
                              0.11980815 0.63584001
## 174
        AMBASLAT data1[, 26]
                               0.08361112 0.74151822
## 175
          AMLAT data1[, 26]
                              0.20436074 0.41597587
## 176
           ENTORH data1[, 26] -0.01583276 0.95028112
           PERIRH data1[, 26]
## 177
                              0.12335004 0.62581283
## 178
              CA1 data1[, 26] -0.06897060 0.78567310
## 179
              CA3 data1[, 26]
                              0.20157377 0.42249904
              DG data1[, 26]
## 180
                               0.12214736 0.62921076
## 181
         CINGULAR data1[, 26]
                               0.26162213 0.29432116
## 182
         PRELIMB data1[, 26]
                               0.52393258 0.02562968
## 183
          SOMSENS data1[, 26]
                               0.39525074 0.10449866
        SUBICULUM data1[, 26]
## 184
                               0.35399941 0.14951909
## 185
          ACCCORE data1[, 26]
                               0.18270271 0.46806182
## 186
         ACCSHELL data1[, 26]
                               0.13056333 0.60558531
## 187
          VISUAL data1[, 26]
                               0.27053288 0.27757543
        PIRIFORM data1[, 26] 0.25667286 0.30388027
## 188
```

```
## 189 PARIETAL data1[, 26] 0.38416871 0.11549134
## 190 RETROSPLEN data1[, 26] 0.35882717 0.14365820
```

First took all the data to see whether they are normally distributed or not: per brain area per group 4 groups, lesion vs. non lesion depending on normality: t test or wilcox

### NORMAL DISTIBUTION?

We will carry out the shapiro-Wilk test. If p>a (bigger than 0.05 generally), the data is normal.

d5h[,c(3,17,20,21)] are not normally distributed d5sh[,c(3)] is not normally distributed d3h[,c(15,21)] are not normally distributed d3sh[,c(3,8,2,14,16,18)] is not normally distributed. You have a small sample size but the population is actually normally distributed, so we will use a parametric test anyway

### ok fuck this shit let's do a t test for all these fuckers

t.test(collumn 1, collumn 2) skip collumns 9:11 for d3h vs d5h

```
for(i in c(3:8,12:21)){
  print(t.test(d3h[,i],d5h[,i]))
}
```

```
##
##
   Welch Two Sample t-test
##
## data: d3h[, i] and d5h[, i]
## t = -1.3064, df = 45.032, p-value = 0.1981
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
  -675.4311 143.9513
## sample estimates:
## mean of x mean of y
##
   1612.441 1878.181
##
##
   Welch Two Sample t-test
##
##
## data: d3h[, i] and d5h[, i]
## t = 1.0996, df = 45.955, p-value = 0.2772
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -115.6838 394.2305
## sample estimates:
## mean of x mean of y
   1339.385 1200.111
##
##
##
   Welch Two Sample t-test
##
##
## data: d3h[, i] and d5h[, i]
## t = -0.32978, df = 38.962, p-value = 0.7433
```

```
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -139.8576 100.6467
## sample estimates:
## mean of x mean of y
## 494.7242 514.3297
##
##
## Welch Two Sample t-test
##
## data: d3h[, i] and d5h[, i]
## t = -0.81178, df = 45.571, p-value = 0.4211
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -323.4522 137.5726
## sample estimates:
## mean of x mean of y
## 1027.418 1120.357
##
##
## Welch Two Sample t-test
## data: d3h[, i] and d5h[, i]
## t = -0.94103, df = 44.666, p-value = 0.3518
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -138.88516
                50.44491
## sample estimates:
## mean of x mean of y
## 324.3175 368.5376
##
##
## Welch Two Sample t-test
##
## data: d3h[, i] and d5h[, i]
## t = -0.50866, df = 44.331, p-value = 0.6135
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -182.7754 109.0943
## sample estimates:
## mean of x mean of y
## 473.4962 510.3367
##
##
## Welch Two Sample t-test
##
## data: d3h[, i] and d5h[, i]
## t = -2.9969, df = 45.89, p-value = 0.004389
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -544.3015 -106.8962
## sample estimates:
## mean of x mean of y
## 1216.220 1541.819
```

```
##
##
##
  Welch Two Sample t-test
##
## data: d3h[, i] and d5h[, i]
## t = -0.70902, df = 45.997, p-value = 0.4819
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -283.2634 135.6917
## sample estimates:
## mean of x mean of y
## 781.9456 855.7315
##
## Welch Two Sample t-test
##
## data: d3h[, i] and d5h[, i]
## t = 1.6793, df = 45.154, p-value = 0.1
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -31.6646 349.4785
## sample estimates:
## mean of x mean of y
## 973.9191 815.0121
##
## Welch Two Sample t-test
## data: d3h[, i] and d5h[, i]
## t = -0.13494, df = 44.752, p-value = 0.8933
\mbox{\tt \#\#} alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -136.0261 118.9459
## sample estimates:
## mean of x mean of y
## 267.2530 275.7931
##
##
## Welch Two Sample t-test
##
## data: d3h[, i] and d5h[, i]
## t = 0.029937, df = 45.939, p-value = 0.9762
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -202.1064 208.2086
## sample estimates:
## mean of x mean of y
## 620.3324 617.2813
##
##
## Welch Two Sample t-test
## data: d3h[, i] and d5h[, i]
## t = 0.85777, df = 45.962, p-value = 0.3955
```

```
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -98.57526 244.96806
## sample estimates:
## mean of x mean of y
## 436.4552 363.2588
##
##
## Welch Two Sample t-test
##
## data: d3h[, i] and d5h[, i]
## t = -0.10222, df = 45.497, p-value = 0.919
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -377.8926 341.3776
## sample estimates:
## mean of x mean of y
  1620.243 1638.500
##
##
## Welch Two Sample t-test
## data: d3h[, i] and d5h[, i]
## t = -0.77948, df = 42.538, p-value = 0.44
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -325.3603 144.0024
## sample estimates:
## mean of x mean of y
## 912.9051 1003.5840
##
##
## Welch Two Sample t-test
##
## data: d3h[, i] and d5h[, i]
## t = 2.282, df = 45.091, p-value = 0.02726
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
    34.44698 552.19530
## sample estimates:
## mean of x mean of y
## 1086.926
             793.605
##
##
## Welch Two Sample t-test
##
## data: d3h[, i] and d5h[, i]
## t = -0.54259, df = 45.789, p-value = 0.59
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -246.8148 142.0159
## sample estimates:
## mean of x mean of y
## 402.9082 455.3077
```

```
for(i in c(3:21)){
  print(t.test(d3sh[,i],d5sh[,i]))
##
## Welch Two Sample t-test
##
## data: d3sh[, i] and d5sh[, i]
## t = -2.8767, df = 27.78, p-value = 0.007635
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -997.2579 -167.5475
## sample estimates:
## mean of x mean of y
## 798.6822 1381.0849
##
##
## Welch Two Sample t-test
## data: d3sh[, i] and d5sh[, i]
## t = -2.1563, df = 31.541, p-value = 0.0388
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -647.7289 -18.2504
## sample estimates:
## mean of x mean of y
## 1031.708 1364.697
##
## Welch Two Sample t-test
##
## data: d3sh[, i] and d5sh[, i]
## t = 0.22244, df = 31.349, p-value = 0.8254
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -119.5437 148.8271
## sample estimates:
## mean of x mean of y
## 575.9308 561.2891
##
##
## Welch Two Sample t-test
##
## data: d3sh[, i] and d5sh[, i]
## t = -2.7633, df = 29.22, p-value = 0.0098
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -602.5997 -90.0839
## sample estimates:
## mean of x mean of y
## 1081.114 1427.455
##
##
```

```
## Welch Two Sample t-test
##
## data: d3sh[, i] and d5sh[, i]
## t = -2.8569, df = 21.807, p-value = 0.009218
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -366.34520 -58.08466
## sample estimates:
## mean of x mean of y
## 312.2207 524.4357
##
##
## Welch Two Sample t-test
##
## data: d3sh[, i] and d5sh[, i]
## t = -3.7813, df = 25.461, p-value = 0.0008469
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -644.1981 -190.1669
## sample estimates:
## mean of x mean of y
## 464.6539 881.8364
##
##
## Welch Two Sample t-test
## data: d3sh[, i] and d5sh[, i]
## t = -4.662, df = 39.474, p-value = 3.537e-05
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -899.4325 -355.2715
## sample estimates:
## mean of x mean of y
## 1065.894 1693.246
##
##
## Welch Two Sample t-test
##
## data: d3sh[, i] and d5sh[, i]
## t = -5.5277, df = 24.487, p-value = 1.027e-05
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -608.6800 -277.9728
## sample estimates:
## mean of x mean of y
## 304.2655 747.5919
##
##
## Welch Two Sample t-test
## data: d3sh[, i] and d5sh[, i]
## t = -1.1798, df = 38.161, p-value = 0.2454
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
```

```
## -140.84168 37.11607
## sample estimates:
## mean of x mean of y
## 256.5610 308.4238
##
## Welch Two Sample t-test
##
## data: d3sh[, i] and d5sh[, i]
## t = -0.59702, df = 23.292, p-value = 0.5563
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -402.6953 222.2184
## sample estimates:
## mean of x mean of y
## 1031.653 1121.892
##
##
##
  Welch Two Sample t-test
## data: d3sh[, i] and d5sh[, i]
## t = -0.44025, df = 31.991, p-value = 0.6627
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -349.0704 224.9968
## sample estimates:
## mean of x mean of y
## 929.2108 991.2476
##
## Welch Two Sample t-test
##
## data: d3sh[, i] and d5sh[, i]
## t = -0.10443, df = 26.241, p-value = 0.9176
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -309.8448 279.8704
## sample estimates:
## mean of x mean of y
## 1018.052 1033.039
##
##
## Welch Two Sample t-test
##
## data: d3sh[, i] and d5sh[, i]
## t = -1.7484, df = 32.93, p-value = 0.08971
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -365.88103
                27.68379
## sample estimates:
## mean of x mean of y
## 505.7771 674.8757
##
##
```

```
## Welch Two Sample t-test
##
## data: d3sh[, i] and d5sh[, i]
## t = -1.9937, df = 29.799, p-value = 0.05541
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -634.816975
                  7.733769
## sample estimates:
## mean of x mean of y
## 633.9459 947.4875
##
##
## Welch Two Sample t-test
##
## data: d3sh[, i] and d5sh[, i]
## t = -0.91218, df = 28.386, p-value = 0.3694
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -421.5918 161.6901
## sample estimates:
## mean of x mean of y
## 547.3349 677.2858
##
##
## Welch Two Sample t-test
## data: d3sh[, i] and d5sh[, i]
## t = -2.482, df = 22.529, p-value = 0.02098
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -1051.34862
                 -94.89004
## sample estimates:
## mean of x mean of y
## 1594.843 2167.963
##
##
## Welch Two Sample t-test
##
## data: d3sh[, i] and d5sh[, i]
## t = -0.1828, df = 29.808, p-value = 0.8562
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -301.8090 252.2322
## sample estimates:
## mean of x mean of y
## 906.9945 931.7829
##
##
## Welch Two Sample t-test
## data: d3sh[, i] and d5sh[, i]
## t = -1.7383, df = 23.722, p-value = 0.09512
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
```

```
## -744.10838 63.94894
## sample estimates:
## mean of x mean of y
## 1136.431 1476.511
##
##
## Welch Two Sample t-test
##
## data: d3sh[, i] and d5sh[, i]
## t = -1.0525, df = 21.17, p-value = 0.3044
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -383.0276 125.5168
## sample estimates:
## mean of x mean of y
## 869.4037 998.1591
```

### FIGURETIME.

```
##
## Attaching package: 'reshape'
##
## The following objects are masked from 'package:reshape2':
##
## colsplit, melt, recast
##
## The following object is masked from 'package:Matrix':
##
## expand
```

