## Circadian Analysis

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## Analysis for activity

```
wt <- read_xlsx("data/Activity_analysis_4_R.xlsx", sheet = 1)</pre>
mut <- read_xlsx("data/Activity_analysis_4_R.xlsx", sheet = 2)</pre>
wt <- wt %>%
  bind_cols(WT.M=rep("WT", nrow(wt)), time = decimal_date(ymd(wt$`Total_revolutions/day`)), .) %>%
  gather(mice, activity, -c(1:5)) %>%
  mutate(time = time-min(time)) %>%
  select(-`Total_revolutions/day`)
mut <- mut %>%
  bind_cols(WT.M=rep("M", nrow(mut)), time = decimal_date(ymd(mut$`Total_revolutions/day`)), .) %>%
  gather(mice, activity, -c(1:5)) %>%
  mutate(time = time-min(time)) %>%
  select(-`Total_revolutions/day`)
data <- wt %>% bind_rows(mut)
data <- data %>% filter(week>=3)
data$mice <- factor(data$mice, levels= unique(data$mice))</pre>
data$time_scaled <- scale(data$time, scale=FALSE)</pre>
data$period <- factor(data$period, levels= unique(data$period))</pre>
data$WT.M <-factor(data$WT.M, levels=c("WT", "M"))</pre>
mod <- lme(activity ~ time_scaled * WT.M, random=~1 mice, data = data)
cat("Estimates, errors and the significance")
## Estimates, errors and the significance
summary(mod)
## Linear mixed-effects model fit by REML
## Data: data
##
          AIC
                   BIC
                          logLik
     8681.339 8705.303 -4334.67
##
## Random effects:
## Formula: ~1 | mice
```

```
(Intercept) Residual
## StdDev:
              14936.81 11161.46
## Fixed effects: activity ~ time_scaled * WT.M
##
                          Value Std.Error DF
                                               t-value p-value
## (Intercept)
                       38778.45
                                  5335.29 388 7.268296 0.0000
## time scaled
                      -20486.52 35588.68 388 -0.575647 0.5652
                                  7810.06 13 -2.602317 0.0219
## WT.MM
                      -20324.26
## time_scaled:WT.MM -289880.69 52096.49 388 -5.564304 0.0000
## Correlation:
##
                     (Intr) tm_scl WT.MM
## time_scaled
                      0.000
## WT.MM
                     -0.683 0.000
## time_scaled:WT.MM 0.000 -0.683 0.000
## Standardized Within-Group Residuals:
##
           Min
                        Q1
                                   Med
                                                QЗ
## -2.69133816 -0.63470177 0.03277689 0.63109234 3.19701990
##
## Number of Observations: 405
## Number of Groups: 15
cat("Bootstrap confidence intervals for the estimates")
## Bootstrap confidence intervals for the estimates
mod_lmer <- lmer(activity ~ time_scaled * WT.M + (1|mice), data = data)</pre>
confint.merMod(mod_lmer, method = "boot", nsim = 999)
## Computing bootstrap confidence intervals ...
##
                           2.5 %
                                      97.5 %
## .sig01
                        8643.361
                                   20403.180
## .sigma
                       10346.041
                                   11960.872
## (Intercept)
                       28744.669
                                   50179.240
## time_scaled
                      -93796.814
                                   44322.642
## WT.MM
                      -36820.325
                                   -5543.353
## time scaled:WT.MM -390349.398 -187440.852
cat("ANOVA table")
## ANOVA table
anova.lme(mod, type = "marginal", adjustSigma = F)
##
                    numDF denDF F-value p-value
## (Intercept)
                            388 52.82812 <.0001
                        1
## time_scaled
                        1
                            388 0.33137 0.5652
## WT.M
                            13 6.77206 0.0219
                        1
## time_scaled:WT.M
                            388 30.96148 <.0001
                        1
```

## Analysis for alpha

```
wt <- read_xlsx("data/Alpha_Activity_analysis_4_R.xlsx", sheet = 1, na = "NA")
mut <- read_xlsx("data/Alpha_Activity_analysis_4_R.xlsx", sheet = 2, na = "NA")</pre>
```

```
wt <- wt %>%
  bind_cols(WT.M = rep("WT", nrow(wt)), time = decimal_date(ymd(wt$`Total_revolutions/day`)), .)%>%
  gather(mice, alpha, -c(1:5)) %>%
  mutate(time = time-min(time)) %>%
  select(-`Total_revolutions/day`)
mut <- mut %>%
  bind cols(WT.M=rep("M", nrow(mut)), time = decimal date(ymd(mut$`Total revolutions/day`)), .)%>%
  gather(mice, alpha, -c(1:5)) %>%
  mutate(time = time-min(time)) %>%
  select(-`Total_revolutions/day`)
alpha_data <- wt %>% bind_rows(mut)
alpha_data <- alpha_data %>% filter(week>=3)
alpha_data <- na.omit(alpha_data)
alpha_data$mice <- factor(alpha_data$mice, levels= unique(alpha_data$mice))</pre>
alpha_data$time_scaled <- scale(alpha_data$time, scale=FALSE)</pre>
alpha_data$period <- factor(alpha_data$period, levels= unique(alpha_data$period))</pre>
alpha data$WT.M <- factor(alpha data$WT.M, levels=c("WT", "M"))
alpha_data$alpha <- as.numeric(alpha_data$alpha)</pre>
mod1 <- lme(alpha ~ time_scaled * WT.M, random=~1 mice, data = alpha_data, na.action = na.omit)
cat("Estimates, errors and the significance")
## Estimates, errors and the significance
summary(mod1)
## Linear mixed-effects model fit by REML
## Data: alpha_data
          AIC
##
                   BIC
                          logLik
     2068.243 2091.978 -1028.121
##
## Random effects:
## Formula: ~1 | mice
           (Intercept) Residual
           0.6720236 3.405597
## StdDev:
##
## Fixed effects: alpha ~ time_scaled * WT.M
                          Value Std.Error DF
                                               t-value p-value
## (Intercept)
                      10.101010 0.334981 373 30.154013 0.0000
## time scaled
                     -16.267322 11.031558 373 -1.474617 0.1412
## WT.MM
                      -0.714526  0.490361  13 -1.457142  0.1688
                     2.360361 16.148547 373 0.146166 0.8839
## time_scaled:WT.MM
## Correlation:
##
                     (Intr) tm_scl WT.MM
## time_scaled
                      0.000
                     -0.683 0.000
## WT.MM
## time_scaled:WT.MM 0.000 -0.683 0.000
## Standardized Within-Group Residuals:
```

```
## -2.73631345 -0.48276146 0.06646042 0.49709145 3.94949030
## Number of Observations: 390
## Number of Groups: 15
cat("Bootstrap confidence intervals for the estimates")
## Bootstrap confidence intervals for the estimates
mod1_lmer <- lmer(alpha ~ time_scaled * WT.M + (1|mice), data = alpha_data)
confint.merMod(mod1_lmer, method = "boot", nsim = 999)
## Computing bootstrap confidence intervals ...
##
                             2.5 %
                                      97.5 %
## .sig01
                     5.386829e-24 1.1201880
                     3.148692e+00 3.6392359
## .sigma
## (Intercept)
                     9.436860e+00 10.7416646
                    -3.886561e+01 7.3806537
## time scaled
## WT.MM
                    -1.622018e+00 0.2386222
## time_scaled:WT.MM -3.048681e+01 34.6770024
cat("ANOVA table")
## ANOVA table
anova.lme(mod1, type = "marginal", adjustSigma = F)
                   numDF denDF F-value p-value
## (Intercept)
                           373 909.2645 <.0001
                       1
## time_scaled
                           373
                                 2.1745 0.1412
                       1
## WT.M
                                 2.1233 0.1688
                       1
                           13
## time scaled:WT.M
                           373
                                0.0214 0.8839
Analysis for period
wt <- read_xlsx("data/Period_analysis_4_R.xlsx", sheet = 1) %>% gather(mice, value, -1)
wt <- data.frame(WT.M=rep("WT", nrow(wt))) %>% bind_cols(wt)
```

```
##
    309.1875 328.7 -144.5938
##
## Random effects:
## Formula: ~1 | mice
          (Intercept) Residual
## StdDev:
           0.7251103 3.268825
## Fixed effects: value ~ week * WT.M
                           Value Std.Error DF
                                              t-value p-value
## (Intercept)
                       23.721429 1.265532 39 18.744234 0.0000
## weekDD_Week_2
                       -3.381429 1.747260 39 -1.935275 0.0602
                        2.055714 1.747260 39 1.176536 0.2465
## weekDD_Week_3
## weekLD_Week_3
                        0.208571 1.747260 39 0.119371 0.9056
## WT.MWT
                        0.137321 1.732901 13 0.079244 0.9380
## weekDD_Week_2:WT.MWT 3.251429 2.392535 39 1.358989 0.1820
## weekDD_Week_3:WT.MWT -2.426964 2.392535 39 -1.014390 0.3166
## weekLD_Week_3:WT.MWT -0.063571 2.392535 39 -0.026571 0.9789
## Correlation:
##
                       (Intr) wkDD_W_2 wkDD_W_3 wkLD_W_3 WT.MWT wDD_W_2:
## weekDD Week 2
                       -0.690
                       -0.690 0.500
## weekDD_Week_3
## weekLD Week 3
                       -0.690 0.500
                                       0.500
## WT.MWT
                       -0.730 0.504
                                      0.504
                                               0.504
## weekDD_Week_2:WT.MWT 0.504 -0.730
                                       -0.365
                                               -0.365
                                                         -0.690
## weekDD_Week_3:WT.MWT 0.504 -0.365
                                      -0.730 -0.365
                                                         -0.690 0.500
## weekLD_Week_3:WT.MWT 0.504 -0.365
                                      -0.365 -0.730
                                                         -0.690 0.500
##
                       wDD_W_3:
## weekDD_Week_2
## weekDD_Week_3
## weekLD_Week_3
## WT.MWT
## weekDD_Week_2:WT.MWT
## weekDD_Week_3:WT.MWT
## weekLD_Week_3:WT.MWT 0.500
## Standardized Within-Group Residuals:
                         Q1
                                     Med
## -5.938352181 -0.067963537 -0.001414478 0.093674882 1.778532447
## Number of Observations: 60
## Number of Groups: 15
cat("Bootstrap confidence intervals for the estimates")
## Bootstrap confidence intervals for the estimates
mod2 lmer <- lmer(value ~ week * WT.M + (1 mice), data = period data)
confint.merMod(mod2_lmer, method = "boot", nsim = 999)
## Computing bootstrap confidence intervals ...
##
                           2.5 %
                                      97.5 %
## .sig01
                        0.000000 2.08600463
## .sigma
                        2.457618 3.79672236
## (Intercept)
                       21.350624 26.08873235
## weekDD Week 2
                       -6.829508 0.06209273
```