Sleep/Wake Analysis

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```
library(tidyverse)
library(lme4)
xl <- read.csv("../data/WT_McKO_combined_sleep.csv")</pre>
colnames(xl)[1] <- gsub('^...','',colnames(xl)[1])</pre>
t_wt <- x1 %>%
   group_by(Transmitter, Genotype, ZT, Day) %>%
   dplyr::summarise(sum_wake = sum(Wake),
                    sum_sws = sum(SWS),
                    sum_ps = sum(PS),
                    sum_art = sum(Artifact))
## 'summarise()' has grouped output by 'Transmitter', 'Genotype', 'ZT'. You can override using the '.gr
t_wt$PercentWake <- (t_wt$sum_wake / (t_wt$sum_sws + t_wt$sum_wake + t_wt$sum_ps + t_wt$sum_art))
t_wt$PercentSWS <- (t_wt$sum_sws / (t_wt$sum_sws + t_wt$sum_wake + t_wt$sum_ps + t_wt$sum_art))</pre>
t_wt$PercentPS <- (t_wt$sum_ps / (t_wt$sum_sws + t_wt$sum_wake + t_wt$sum_ps + t_wt$sum_art))
t_wt$PercentArt <- ((t_wt$sum_art)/ (t_wt$sum_sws + t_wt$sum_wake + t_wt$sum_ps + t_wt$sum_art))
t_wt <- t_wt %>% dplyr::mutate(phase= ifelse(ZT>11, "dark", "light"))
t_wt$ZT <- as.factor(t_wt$ZT)</pre>
SumStatwt <- dplyr::summarise(group_by(t_wt, Genotype, ZT, phase),</pre>
n = n(),
mean_Wake = mean(PercentWake),
sd_Wake = sd(PercentWake),
se_Wake = sd_Wake/sqrt(n),
mean SWS = mean(PercentSWS),
sd_SWS = sd(PercentSWS),
se_SWS = sd_SWS/sqrt(n),
mean_PS = mean(PercentPS),
sd_PS = sd(PercentPS),
se_PS = sd_PS/sqrt(n))
```

'summarise()' has grouped output by 'Genotype', 'ZT'. You can override using the '.groups' argument.

```
## # A tibble: 48 x 13
## # Groups:
             Genotype, ZT [48]
##
     Genotype ZT
                   phase
                            n mean_Wake sd_Wake se_Wake mean_SWS sd_SWS se_SWS
##
             <fct> <chr> <int>
                                                <dbl>
                                                         <dbl> <dbl> <dbl>
     <chr>
                                  <dbl>
                                         <dbl>
                                         0.140 0.0222
## 1 McKO
             0
                   light
                           40
                                  0.207
                                                         0.720 0.141 0.0223
## 2 McKO
                                 0.222  0.173  0.0274  0.709  0.162  0.0257
             1
                   light
                           40
                                                        0.767 0.147 0.0232
## 3 McKO
             2
                   light
                           40
                                 0.158
                                         0.148 0.0234
## 4 McKO
             3
                           40
                                 0.206  0.139  0.0219  0.716  0.128  0.0203
                   light
## 5 McKO
                           40
                                 4
                   light
                                                      0.726 0.176 0.0279
## 6 McKO
             5
                   light
                           40
                                 0.209
                                         0.185 0.0292
## 7 McKO
             6
                                 0.179
                                        0.155 0.0244
                                                        0.750 0.149 0.0235
                   light
                           40
## 8 McKO
             7
                   light
                           40
                                 0.175  0.160  0.0253  0.748  0.159  0.0251
## 9 McKO
             8
                           40
                                 0.197
                                         0.141 0.0224
                                                        0.724 0.132 0.0208
                   light
## 10 McKO
             9
                   light
                           40
                                  0.276
                                         0.178 0.0281
                                                         0.654 0.171 0.0270
## # ... with 38 more rows, and 3 more variables: mean_PS <dbl>, sd_PS <dbl>,
## # se_PS <dbl>
t_wt_days <- t_wt %>%
  group_by(Transmitter, Genotype, ZT, phase) %>%
  dplyr::summarise(mean wake = mean(PercentWake),
                  mean_sws = mean(PercentSWS),
                  mean_ps = mean(PercentPS))
```

'summarise()' has grouped output by 'Transmitter', 'Genotype', 'ZT'. You can override using the '.gr

```
SumStatwt$ZT <- as.numeric(as.character(SumStatwt$ZT))</pre>
wake_plot <- ggplot(data=SumStatwt, aes(ZT, mean_Wake, color=Genotype)) +</pre>
  geom_line(aes(group=Genotype)) +
  geom_point(stat="identity", position=position_dodge(), size=2) +
  geom_errorbar(data=SumStatwt,
                aes(x=ZT, ymin=mean Wake-se Wake,
                    ymax=mean_Wake+se_Wake, color=Genotype),
                width=.5) +
  scale_y_continuous(labels = scales::percent,
                     name = "Average % Time Spent Awake",
                     limits = c(0.1, 0.9)) +
  scale_x_continuous(limits=c(-1,24), breaks = seq(from=0, to=23, by=2)) +
  annotate("rect", xmin=12, xmax=23, ymin=-Inf, ymax=Inf,
           alpha=0.2, fill="black") +
  theme_bw()
#qqsave("../figures/wake_plot.svq", plot=wake_plot, height=4, width=7)
sleep_plot <- ggplot(data=SumStatwt, aes(ZT, mean_Sleep, color=Genotype)) +</pre>
  geom_line(aes(group=Genotype)) +
  geom_point(stat="identity", position=position_dodge(), size=2) +
  geom_errorbar(data=SumStatwt,
                aes(x=ZT, ymin=mean Sleep-se Sleep,
                    ymax=mean_Sleep+se_Sleep, color=Genotype),
```

```
width=.5) +
  scale_y_continuous(labels = scales::percent,
                     name = "Average % Time Spent Asleep",
                     limits = c(0.15, 0.9)) +
  scale_x_continuous(limits=c(-1,24), breaks = seq(from=0, to=23, by=2)) +
  annotate("rect", xmin=12, xmax=23, ymin=-Inf, ymax=Inf,
           alpha=0.2, fill="black") +
  theme bw()
#ggsave("../figures/sleep_plot.svg", plot=sleep_plot, height=4, width=7)
sws_plot <- ggplot(data=SumStatwt, aes(ZT, mean_SWS, color=Genotype)) +</pre>
  geom line(aes(group=Genotype)) +
  geom_point(stat="identity", position=position_dodge(), size=2) +
  geom_errorbar(data=SumStatwt,
                aes(x=ZT, ymin=mean_SWS-se_SWS,
                    ymax=mean_SWS+se_SWS, color=Genotype),
                width=.5) +
  scale_y_continuous(labels = scales::percent,
                     name = "Average % Time Spent in SWS",
                     limits = c(0.1, 0.85)) +
  scale_x_continuous(limits=c(-1,24), breaks = seq(from=0, to=23, by=2)) +
  annotate("rect", xmin=12, xmax=23, ymin=-Inf, ymax=Inf,
           alpha=0.2, fill="black") +
  theme bw()
#ggsave("../figures/sws_plot.svg", plot=sws_plot, height=4, width=7)
ps_plot <- ggplot(data=SumStatwt, aes(ZT, mean_PS, color=Genotype)) +</pre>
  geom_line(aes(group=Genotype)) +
  geom_point(stat="identity", position=position_dodge(), size=2) +
  geom_errorbar(data=SumStatwt,
                aes(x=ZT, ymin=mean_PS-se_PS,
                    ymax=mean_PS+se_PS, color=Genotype),
                width=.5) +
  scale_y_continuous(labels = scales::percent,
                     name = "Average % Time Spent in Paradoxical Sleep",
                     limits = c(0.02, 0.1) +
  scale_x_continuous(limits=c(-1,24), breaks = seq(from=0, to=23, by=2)) +
  annotate("rect", xmin=12, xmax=23, ymin=-Inf, ymax=Inf,
           alpha=0.2, fill="black") +
  theme_bw()
#ggsave("../figures/ps_plot.png", plot=ps_plot, height=4, width=7)
SumStatwtp <- dplyr::summarise(group_by(t_wt, Genotype, phase),</pre>
n = n(),
mean_Wake = mean(PercentWake),
sd_Wake = sd(PercentWake),
se_Wake = sd_Wake/sqrt(n),
mean_SWS = mean(PercentSWS),
sd_SWS = sd(PercentSWS),
se_SWS = sd_SWS/sqrt(n),
```

```
mean_PS = mean(PercentPS),
sd_PS = sd(PercentPS),
se_PS = sd_PS/sqrt(n))
## 'summarise()' has grouped output by 'Genotype'. You can override using the '.groups' argument.
SumStatwtp
## # A tibble: 4 x 12
## # Groups: Genotype [2]
    Genotype phase n mean_Wake sd_Wake se_Wake mean_SWS sd_SWS se_SWS mean_PS
                                                                             <dbl>
##
    <chr>
             <chr> <int>
                             <dbl> <dbl> <dbl>
                                                      <dbl> <dbl> <dbl>
## 1 McKO
             dark
                     480
                             0.521
                                     0.223 0.0102
                                                      0.437 0.211 0.00962 0.0395
## 2 McKO
                             0.229 0.180 0.00824
                                                      0.700 0.173 0.00790 0.0689
             light
                     480
## 3 WT
                     552
                             0.558
                                   0.279 0.0119
                                                      0.395 0.267 0.0114
             dark
                                                                            0.0437
                             0.270 0.201 0.00854
                                                      0.653 0.190 0.00808 0.0745
## 4 WT
             light 552
## # ... with 2 more variables: sd_PS <dbl>, se_PS <dbl>
t_wtp <- t_wt %>%
 group_by(Transmitter, Genotype, phase)%>%
 dplyr::summarise(mean_wake = mean(PercentWake),
                  mean_sws = mean(PercentSWS),
                  mean ps = mean(PercentPS),
                  mean_art = mean(PercentArt))
## 'summarise()' has grouped output by 'Transmitter', 'Genotype'. You can override using the '.groups'
phase_sumstats <- t_wtp %>% group_by(Genotype, phase) %>%
 summarise(n = n(),
           mean_wake_perc = mean(mean_wake),
           sd_wake_perc = sd(mean_wake),
           mean_sws_perc = mean(mean_sws),
           sd_sws_perc = sd(mean_sws),
           mean_ps_perc = mean(mean_ps),
           sd_ps_perc = sd(mean_ps))
## 'summarise()' has grouped output by 'Genotype'. You can override using the '.groups' argument.
phase_sumstats
## # A tibble: 4 x 9
## # Groups:
              Genotype [2]
##
    Genotype phase
                       n mean_wake_perc sd_wake_perc mean_sws_perc sd_sws_perc
    <chr>
             <chr> <int>
                                <dbl>
                                              <dbl>
                                                             <dbl>
                                                                         <dbl>
                                  0.546
                                              0.0700
                                                             0.414
                                                                        0.0696
## 1 McKO
             dark
                       9
## 2 McKO
             light
                       9
                                  0.224
                                              0.0349
                                                             0.704
                                                                        0.0578
## 3 WT
             dark
                      10
                                  0.560
                                              0.0411
                                                             0.392
                                                                        0.0555
## 4 WT
             light
                      10
                                  0.268
                                              0.0360
                                                             0.655
                                                                        0.0696
```

... with 2 more variables: mean_ps_perc <dbl>, sd_ps_perc <dbl>

```
temp <- t_wt %>%
      group_by(Transmitter, Genotype)%>%
      dplyr::summarise(mean_wake = mean(PercentWake),
                                                             mean_sws = mean(PercentSWS),
                                                             mean_ps = mean(PercentPS),
                                                             mean_Art = mean(PercentArt))
## 'summarise()' has grouped output by 'Transmitter'. You can override using the '.groups' argument.
max(temp$mean_Art)
## [1] 0.006751485
temp %>% group_by(Genotype) %>%
      summarise(n = n(),
                                      wake_perc = mean(mean_wake),
                                      sd_wake = sd(mean_wake),
                                      sws_perc = mean(mean_sws),
                                      sd_sws = sd(mean_sws),
                                       ps_perc = mean(mean_ps),
                                      sd_ps = sd(mean_ps))
## # A tibble: 2 x 8
               Genotype n wake_perc sd_wake sws_perc sd_sws ps_perc sd_ps
                <chr> <int>
                                                                             <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <
## 1 McKO
                                                     9
                                                                             0.385 0.0370
                                                                                                                             0.559 0.0497 0.0544 0.0233
## 2 WT
                                                                             0.414 0.0327 0.524 0.0568 0.0590 0.0291
                                                      10
#ggplot() +
\# \ geom\_bar(data=SumStatwtp,\ aes(phase,\ mean\_Wake,fill=Genotype),\ stat="identity",\ position=position\_dotatwoodly and the property of th
\# qeom_errorbar(data=SumStatwtp, aes(x=phase,ymin=mean_Wake - se_Wake, ymax=mean_Wake + se_Wake, fill=
\# geom\_point(data=t\_wtp, aes(x=phase, y=mean\_wake, fill=Genotype), position=position\_dodge(width=0.5))
#ggplot() +
\# geom\_bar(data=SumStatwtp, aes(phase, mean\_Sleep,fill=Genotype), stat="identity", position=position\_data=SumStatwtp, aes(phase, mean\_Sleep,fill=Genotype), stat="identity", position=position_data=SumStatwtp, aes(phase, mean\_Sleep,fill=Genotype), stat="identity", position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position=position
\# geom_errorbar(data=SumStatwtp, aes(x=phase,ymin=mean_Sleep - se_Sleep, ymax=mean_Sleep + se_Sleep, f
\# geom_point(data=t_wtp, aes(x=phase, y=mean_sleep, fill=Genotype), position=position_dodqe(width=0.5)
t_wt$ZT <- as.factor(t_wt$ZT)</pre>
t_wt$phase <- as.factor(t_wt$phase)</pre>
t_wt$Genotype <- as.factor(t_wt$Genotype)</pre>
t_wt_days$phase <- as.factor(t_wt_days$phase)</pre>
t_wt_days$Genotype <- as.factor(t_wt_days$Genotype)</pre>
t_wt_days$Transmitter <- as.factor(t_wt_days$Transmitter)</pre>
#over entire day
wake_day <- wilcox.test(mean_wake ~ Genotype, data= t_wt_days)</pre>
wake_day
```

```
##
## Wilcoxon rank sum test with continuity correction
## data: mean_wake by Genotype
## W = 24278, p-value = 0.2427
## alternative hypothesis: true location shift is not equal to 0
sws_day <- wilcox.test(mean_sws ~ Genotype, data=t_wt_days)</pre>
sws_day
##
## Wilcoxon rank sum test with continuity correction
## data: mean_sws by Genotype
## W = 27970, p-value = 0.1447
## alternative hypothesis: true location shift is not equal to 0
ps_day <- wilcox.test(mean_ps ~ Genotype, data=t_wt_days)</pre>
ps_day
##
## Wilcoxon rank sum test with continuity correction
##
## data: mean_ps by Genotype
## W = 24881, p-value = 0.4598
## alternative hypothesis: true location shift is not equal to 0
#by phase
lmer_wake <- lmer(mean_wake ~ phase*Genotype+(1|Transmitter), data = t_wt_days)</pre>
#plot(lmer_wake, type=c("p", "smooth"), col.line=1)
#lattice::qqmath(lmer_wake)
anova(lmer_wake)
## Analysis of Variance Table
                 npar Sum Sq Mean Sq F value
##
## phase
                    1 10.7003 10.7003 432.2658
## Genotype
                    1 0.0838 0.0838
                                       3.3843
                    1 0.0253 0.0253 1.0225
## phase:Genotype
emmeans::emmeans(lmer_wake, pairwise ~ Genotype | phase)$contrasts
## phase = dark:
## contrast estimate SE df t.ratio p.value
## McKO - WT -0.0145 0.0218 56.7 -0.666 0.5080
## phase = light:
## contrast estimate SE df t.ratio p.value
## McKO - WT -0.0443 0.0218 56.7 -2.038 0.0463
## Degrees-of-freedom method: kenward-roger
```

```
emmeans::emmeans(lmer_wake, pairwise ~ phase | Genotype)$contrasts
## Genotype = McKO:
## contrast estimate
                            SE df t.ratio p.value
## dark - light 0.322 0.0214 435 15.043 <.0001
##
## Genotype = WT:
## contrast estimate SE df t.ratio p.value
## dark - light 0.292 0.0203 435 14.387 <.0001
##
## Degrees-of-freedom method: kenward-roger
lmer_sws <- lmer(mean_sws ~ phase*Genotype+(1|Transmitter), data = t_wt_days)</pre>
#plot(lmer_sws, type=c("p","smooth"), col.line=1)
#lattice::qqmath(lmer_sws)
anova(lmer_sws)
## Analysis of Variance Table
##
                npar Sum Sq Mean Sq F value
## phase
                  1 8.6962 8.6962 405.4033
                   1 0.0442 0.0442 2.0607
## Genotype
## phase:Genotype 1 0.0208 0.0208 0.9675
emmeans::emmeans(lmer_sws, pairwise ~ Genotype | phase)$contrasts
## phase = dark:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.0218 0.0282 29.1 0.774 0.4451
##
## phase = light:
## contrast estimate
                        SE df t.ratio p.value
## McKO - WT 0.0488 0.0282 29.1 1.733 0.0937
##
## Degrees-of-freedom method: kenward-roger
emmeans::emmeans(lmer_sws, pairwise ~ phase | Genotype)$contrasts
## Genotype = McKO:
## contrast estimate
                            SE df t.ratio p.value
## dark - light -0.290 0.0199 435 -14.571 <.0001
##
## Genotype = WT:
## contrast estimate SE df t.ratio p.value
## dark - light -0.263 0.0189 435 -13.930 <.0001
##
## Degrees-of-freedom method: kenward-roger
lmer_ps <- lmer(mean_ps ~ phase*Genotype+(1|Transmitter), data = t_wt_days)</pre>
#plot(lmer_ps, type=c("p", "smooth"), col.line=1)
#lattice::gqmath(lmer ps)
anova(lmer_ps)
```

```
## Analysis of Variance Table
##
                 npar Sum Sq Mean Sq F value
                    1 0.115184 0.115184 178.5871
## phase
                    1 0.000092 0.000092 0.1430
## Genotype
## phase:Genotype
                    1 0.000139 0.000139 0.2159
#emmeans::emmeans(lmer_ps, pairwise ~ Genotype / phase)$contrasts
emmeans::emmeans(lmer_ps, pairwise ~ phase | Genotype)$contrasts
## Genotype = McKO:
## contrast estimate
                              SE df t.ratio p.value
## dark - light -0.0330 0.00346 435 -9.535 <.0001
## Genotype = WT:
                estimate
## contrast
                              SE df t.ratio p.value
## dark - light -0.0307 0.00328 435 -9.375 <.0001
##
## Degrees-of-freedom method: kenward-roger
#by ZT
t_wt_days$ZT <- as.factor(t_wt_days$ZT)</pre>
lmer_wake_zt <- lmer(mean_wake ~ ZT*Genotype+(1|Transmitter), data = t_wt_days)</pre>
#plot(lmer_wake_zt, type=c("p", "smooth"), col.line=1)
#lattice::qqmath(lmer_wake_zt)
anova(lmer_wake_zt)
## Analysis of Variance Table
              npar Sum Sq Mean Sq F value
## ZT
                23 13.4075 0.58293 32.0622
                1 0.0615 0.06153 3.3843
## Genotype
## ZT:Genotype 23 0.9773 0.04249 2.3372
#emmeans::emmeans(lmer_wake_zt, pairwise ~ ZT / Genotype)
emmeans::emmeans(lmer_wake_zt, pairwise ~ Genotype | ZT)$contrasts
## ZT = 0:
## contrast
              estimate
                           SE df t.ratio p.value
## McKO - WT -1.23e-01 0.0627 403 -1.956 0.0511
##
## ZT = 1:
## contrast
              estimate
                           SE df t.ratio p.value
## McKO - WT -7.01e-04 0.0627 403 -0.011 0.9911
##
## ZT = 2:
## contrast
              estimate
                           SE df t.ratio p.value
## McKO - WT -6.66e-02 0.0627 403 -1.062 0.2889
##
## ZT = 3:
## contrast
              estimate
                           SE df t.ratio p.value
## McKO - WT -4.31e-02 0.0627 403 -0.687 0.4924
##
```

```
## ZT = 4:
## contrast estimate SE df t.ratio p.value
## McKO - WT -1.01e-01 0.0627 403 -1.612 0.1077
##
## ZT = 5:
## contrast estimate
                        SE df t.ratio p.value
## McKO - WT 1.00e-02 0.0627 403 0.160 0.8728
##
## ZT = 6:
## contrast
             estimate
                         SE df t.ratio p.value
## McKO - WT 2.54e-03 0.0627 403
                                 0.040 0.9677
##
## ZT = 7:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT -1.28e-01 0.0627 403 -2.040 0.0420
##
## ZT = 8:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT -5.38e-02 0.0627 403 -0.857 0.3917
## ZT = 9:
## contrast estimate
                        SE df t.ratio p.value
## McKO - WT -2.96e-02 0.0627 403 -0.472 0.6372
## ZT = 10:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT 9.08e-03 0.0627 403
                                 0.145 0.8850
## ZT = 11:
                         SE df t.ratio p.value
## contrast estimate
## McKO - WT -8.19e-03 0.0627 403 -0.131 0.8962
##
## ZT = 12:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT 1.18e-01 0.0627 403 1.884 0.0603
## ZT = 13:
## contrast estimate
                        SE df t.ratio p.value
## McKO - WT -8.80e-02 0.0627 403 -1.403 0.1613
##
## ZT = 14:
                         SE df t.ratio p.value
## contrast estimate
## McKO - WT -2.34e-01 0.0627 403 -3.725 0.0002
##
## ZT = 15:
                         SE df t.ratio p.value
## contrast estimate
## McKO - WT -1.65e-01 0.0627 403 -2.632 0.0088
##
## ZT = 16:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT -5.53e-02 0.0627 403 -0.882 0.3781
##
## ZT = 17:
## contrast estimate
                      SE df t.ratio p.value
```

```
## McKO - WT 6.15e-02 0.0627 403 0.981 0.3274
##
## ZT = 18:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT 1.71e-01 0.0627 403
                                  2.722 0.0068
##
## ZT = 19:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT 9.30e-02 0.0627 403
                                   1.483 0.1389
##
## ZT = 20:
                          SE df t.ratio p.value
## contrast
              estimate
## McKO - WT 7.88e-02 0.0627 403
                                  1.256 0.2097
##
## ZT = 21:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT -1.05e-02 0.0627 403 -0.168 0.8666
##
## ZT = 22:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT 5.45e-05 0.0627 403 0.001 0.9993
## ZT = 23:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT -1.44e-01 0.0627 403 -2.289 0.0226
## Degrees-of-freedom method: kenward-roger
lmer_sws_zt <- lmer(mean_sws ~ ZT*Genotype+(1|Transmitter), data = t_wt_days)</pre>
#plot(lmer_sws_zt, type=c("p","smooth"), col.line=1)
#lattice::qqmath(lmer_sws_zt)
anova(lmer_sws_zt)
## Analysis of Variance Table
             npar Sum Sq Mean Sq F value
## ZT
                23 11.0658 0.48112 30.8128
## Genotype
                1 0.0322 0.03218 2.0607
                23 0.8769 0.03813 2.4418
## ZT:Genotype
#emmeans::emmeans(lmer_sws_zt, pairwise ~ ZT | Genotype)
emmeans::emmeans(lmer_sws_zt, pairwise ~ Genotype | ZT)$contrasts
## ZT = 0:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT 0.10913 0.0614 301
                                  1.779 0.0763
##
## ZT = 1:
                         SE df t.ratio p.value
## contrast estimate
## McKO - WT 0.00306 0.0614 301 0.050 0.9602
##
## ZT = 2:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT 0.07281 0.0614 301 1.187 0.2363
```

```
##
## ZT = 3:
## contrast estimate
                        SE df t.ratio p.value
## McKO - WT 0.02681 0.0614 301 0.437 0.6625
## ZT = 4:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.10417 0.0614 301 1.698 0.0906
##
## ZT = 5:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.01395 0.0614 301 0.227 0.8203
## ZT = 6:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.01475 0.0614 301 0.240 0.8102
##
## ZT = 7:
## contrast estimate
                       SE df t.ratio p.value
## McKO - WT 0.12452 0.0614 301 2.029 0.0433
##
## ZT = 8:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.06289 0.0614 301 1.025 0.3062
##
## ZT = 9:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.03490 0.0614 301 0.569 0.5699
##
## ZT = 10:
## contrast estimate
                     SE df t.ratio p.value
## McKO - WT 0.00325 0.0614 301 0.053 0.9578
##
## ZT = 11:
## contrast estimate
                     SE df t.ratio p.value
## McKO - WT 0.01588 0.0614 301 0.259 0.7960
##
## ZT = 12:
## contrast estimate SE df t.ratio p.value
## McKO - WT -0.10121 0.0614 301 -1.650 0.1001
##
## ZT = 13:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.10178 0.0614 301 1.659 0.0982
## ZT = 14:
## contrast estimate
                        SE df t.ratio p.value
## McKO - WT 0.23898 0.0614 301
                                 3.895 0.0001
##
## ZT = 15:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.16625 0.0614 301 2.709 0.0071
##
## ZT = 16:
```

```
SE df t.ratio p.value
## contrast estimate
## McKO - WT 0.05593 0.0614 301 0.911 0.3628
##
## ZT = 17:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT -0.05698 0.0614 301 -0.929 0.3538
## ZT = 18:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT -0.15191 0.0614 301 -2.476 0.0138
##
## ZT = 19:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT -0.08947 0.0614 301 -1.458 0.1459
##
## ZT = 20:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT -0.06989 0.0614 301 -1.139 0.2556
##
## ZT = 21:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT 0.02042 0.0614 301
                                 0.333 0.7395
##
## ZT = 22:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT 0.01968 0.0614 301
                                 0.321 0.7487
##
## ZT = 23:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT 0.12827 0.0614 301
                                  2.091 0.0374
##
## Degrees-of-freedom method: kenward-roger
lmer_ps_zt <- lmer(mean_ps ~ ZT*Genotype+(1|Transmitter), data = t_wt_days)</pre>
#plot(lmer_ps_zt, type=c("p", "smooth"), col.line=1)
#lattice::qqmath(lmer_ps_zt)
anova(lmer_ps_zt)
## Analysis of Variance Table
##
              npar Sum Sq Mean Sq F value
## ZT
                23 0.133665 0.0058115 9.1017
## Genotype
                1 0.000091 0.0000913 0.1430
## ZT:Genotype 23 0.012563 0.0005462 0.8555
#emmeans::emmeans(lmer_ps_zt, pairwise ~ ZT / Genotype)
emmeans::emmeans(lmer_ps_zt, pairwise ~ Genotype | ZT)$contrasts
## ZT = 0:
## contrast
              estimate
                          SE
                               df t.ratio p.value
## McKO - WT 0.014341 0.0167 57.5
                                  0.860 0.3932
##
## ZT = 1:
## contrast estimate
                          SE
                               df t.ratio p.value
```

```
## McKO - WT -0.001925 0.0167 57.5 -0.115 0.9085
##
## ZT = 2:
## contrast estimate SE df t.ratio p.value
## McKO - WT -0.004487 0.0167 57.5 -0.269 0.7888
##
## ZT = 3:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.016226 0.0167 57.5 0.973 0.3345
##
## ZT = 4:
                         SE
## contrast estimate
                             df t.ratio p.value
## McKO - WT -0.001553 0.0167 57.5 -0.093 0.9261
##
## ZT = 5:
## contrast estimate SE df t.ratio p.value
## McKO - WT -0.023187 0.0167 57.5 -1.391 0.1696
##
## ZT = 6:
## contrast estimate SE df t.ratio p.value
## McKO - WT -0.016316 0.0167 57.5 -0.979 0.3318
## ZT = 7:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.005188 0.0167 57.5 0.311 0.7568
## ZT = 8:
                         SE
## contrast estimate
                             df t.ratio p.value
## McKO - WT -0.007859 0.0167 57.5 -0.471 0.6391
##
## ZT = 9:
## contrast estimate
                         SE
                             df t.ratio p.value
## McKO - WT -0.003685 0.0167 57.5 -0.221 0.8258
##
## ZT = 10:
## contrast estimate SE df t.ratio p.value
## McKO - WT -0.011978 0.0167 57.5 -0.719 0.4753
##
## ZT = 11:
## contrast estimate
                         SE
                             df t.ratio p.value
## McKO - WT -0.006817 0.0167 57.5 -0.409 0.6841
##
## ZT = 12:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT -0.017422 0.0167 57.5 -1.045 0.3003
##
## ZT = 13:
## contrast estimate
                         SE
                             df t.ratio p.value
## McKO - WT -0.010501 0.0167 57.5 -0.630 0.5312
##
## ZT = 14:
## contrast estimate SE df t.ratio p.value
## McKO - WT -0.003280 0.0167 57.5 -0.197 0.8447
##
```

```
## ZT = 15:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.000754 0.0167 57.5 0.045 0.9641
##
## ZT = 16:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.001628 0.0167 57.5 0.098 0.9225
##
## ZT = 17:
## contrast estimate
                         SE
                              df t.ratio p.value
## McKO - WT -0.003584 0.0167 57.5 -0.215 0.8305
##
## ZT = 18:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT -0.018485 0.0167 57.5 -1.109 0.2721
##
## ZT = 19:
## contrast estimate SE df t.ratio p.value
## McKO - WT -0.003875 0.0167 57.5 -0.232 0.8170
## ZT = 20:
## contrast estimate SE df t.ratio p.value
## McKO - WT -0.007489 0.0167 57.5 -0.449 0.6549
## ZT = 21:
## contrast estimate
                         SE
                             df t.ratio p.value
## McKO - WT -0.008013 0.0167 57.5 -0.481 0.6326
## ZT = 22:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT -0.017233 0.0167 57.5 -1.034 0.3056
##
## ZT = 23:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.018886 0.0167 57.5 1.133 0.2620
## Degrees-of-freedom method: kenward-roger
```