Sleep_score_CTB

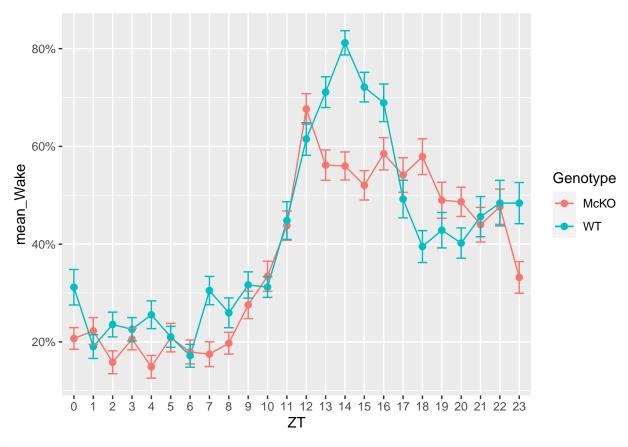
nbergum

3/25/2022

```
library(tidyverse)
library(lme4)
#1 hour bins (by ZT)
xl <- read.csv("../data/McKO_WT_combined.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))</pre>
t_wt$PercentSWS <- (t_wt$sum_sws / (t_wt$sum_sws + t_wt$sum_wake + t_wt$sum_ps + t_wt$sum_art))
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$PercentSleep <- ((t_wt$sum_sws + t_wt$sum_ps)/ (t_wt$sum_sws + t_wt$sum_wake + t_wt$sum_ps + t_wt$
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),
n = n(),
mean_Wake = mean(PercentWake),
sd Wake = sd(PercentWake),
se_Wake = sd_Wake/sqrt(n),
mean_Sleep = mean(PercentSleep),
sd_Sleep = sd(PercentSleep),
se_Sleep = sd_Sleep/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.

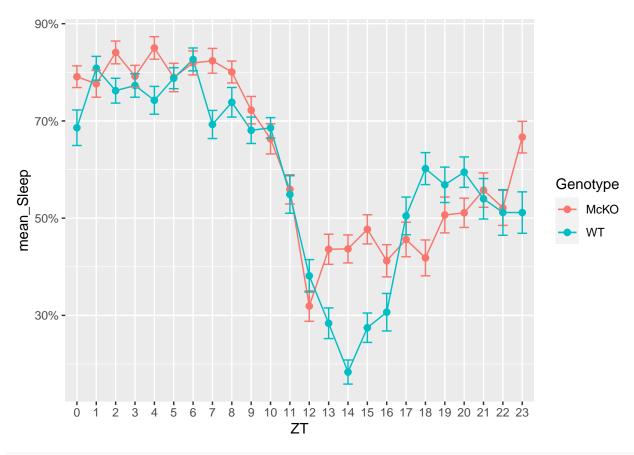
```
SumStatwt
## # A tibble: 48 x 16
              Genotype, ZT [48]
## # Groups:
##
      Genotype ZT
                     phase
                               n mean_Wake sd_Wake se_Wake mean_Sleep sd_Sleep
##
                                                     <dbl>
      <chr>
               <fct> <chr> <int>
                                     <dbl>
                                             <dbl>
                                                                <dbl>
                                                                         <dbl>
##
   1 McKO
                                     0.207
                                             0.140 0.0222
                                                                0.791
                                                                         0.141
              0
                     light
                              40
   2 McKO
                                             0.173 0.0274
                                                                         0.174
##
              1
                     light
                              40
                                     0.222
                                                                0.776
## 3 McKO
              2
                     light
                              40
                                     0.158
                                             0.148 0.0234
                                                                0.841
                                                                         0.148
## 4 McKO
              3
                     light
                              40
                                     0.206
                                             0.139 0.0219
                                                                0.792
                                                                         0.141
## 5 McKO
               4
                     light
                              40
                                     0.149
                                             0.147 0.0232
                                                                0.850
                                                                         0.147
## 6 McKO
               5
                                     0.209
                                             0.185 0.0292
                                                                0.789
                                                                         0.185
                     light
                              40
## 7 McKO
               6
                     light
                              40
                                     0.179
                                             0.155 0.0244
                                                                0.819
                                                                         0.156
## 8 McKO
              7
                                     0.175
                                             0.160 0.0253
                                                                         0.161
                     light
                              40
                                                                0.824
## 9 McKO
               8
                     light
                              40
                                     0.197
                                             0.141 0.0224
                                                                0.801
                                                                         0.143
## 10 McKO
               9
                     light
                              40
                                     0.276
                                             0.178 0.0281
                                                                0.722
                                                                         0.178
## # ... with 38 more rows, and 7 more variables: se_Sleep <dbl>, mean_SWS <dbl>,
## # sd_SWS <dbl>, se_SWS <dbl>, 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),
                    mean_sleep = mean(PercentSleep))
## `summarise()` has grouped output by 'Transmitter', 'Genotype', 'ZT'. You can override using the `.gr
ggplot(data=SumStatwt, aes(ZT, mean_Wake, color=Genotype)) + geom_line(aes(group=Genotype)) + geom_poin
## Warning: Ignoring unknown aesthetics: fill
## Warning: Width not defined. Set with `position_dodge(width = ?)`
```



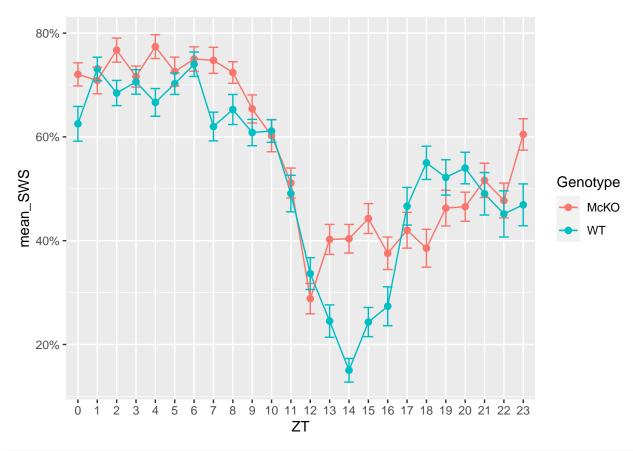
ggplot(data=SumStatwt, aes(ZT, mean_Sleep, color=Genotype)) + geom_line(aes(group=Genotype)) + geom_poi

Warning: Ignoring unknown aesthetics: fill

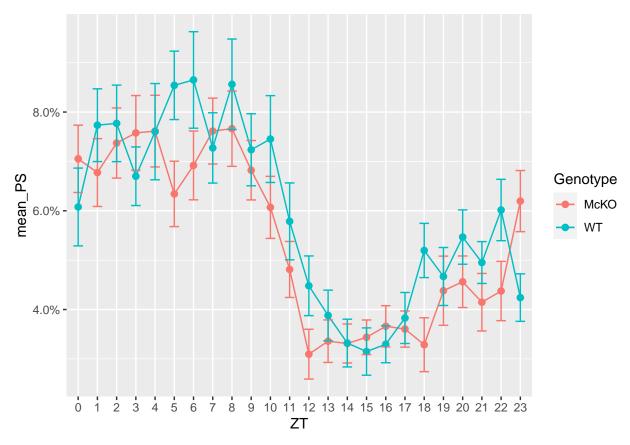
Warning: Width not defined. Set with `position_dodge(width = ?)`



ggplot(data=SumStatwt, aes(ZT, mean_SWS, color=Genotype)) + geom_line(aes(group=Genotype)) + geom_point
Warning: Width not defined. Set with `position_dodge(width = ?)`



ggplot(data=SumStatwt, aes(ZT, mean_PS, color=Genotype)) + geom_line(aes(group=Genotype)) + geom_point(
Warning: Width not defined. Set with `position_dodge(width = ?)`



```
SumStatwtp <- dplyr::summarise(group_by(t_wt, Genotype, phase),
n = n(),
mean_Wake = mean(PercentWake),
sd_Wake = sd(PercentWake),
se_Wake = sd_Wake/sqrt(n),
mean_Sleep = mean(PercentSleep),
sd_Sleep = sd(PercentSleep),
se_Sleep = sd_Sleep/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))</pre>
```

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

```
## # A tibble: 4 x 15
## # Groups:
              Genotype [2]
##
     Genotype phase
                        n mean_Wake sd_Wake se_Wake mean_Sleep sd_Sleep se_Sleep
##
     <chr>
              <chr> <int>
                              <dbl>
                                      <dbl>
                                              <dbl>
                                                         <dbl>
                                                                  <dbl>
                                                                            <dbl>
## 1 McKO
              dark
                      480
                              0.521
                                      0.223 0.0102
                                                         0.477
                                                                  0.224 0.0102
## 2 McKO
                              0.229
                                      0.180 0.00824
                                                         0.769
                                                                  0.181 0.00828
              light
                      480
## 3 WT
              dark
                      552
                              0.558
                                      0.279 0.0119
                                                         0.439
                                                                  0.280 0.0119
## 4 WT
              light
                      552
                              0.270
                                    0.201 0.00854
                                                         0.728
                                                                  0.202 0.00858
```

```
## # ... with 6 more variables: mean_SWS <dbl>, sd_SWS <dbl>, se_SWS <dbl>,
## # mean_PS <dbl>, 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 sleep = mean(PercentSleep))
## `summarise()` has grouped output by 'Transmitter', 'Genotype'. You can override using the `.groups`
# geom_bar(data=SumStatwtp, aes(phase, mean_Wake,fill=Genotype), stat="identity", position=position_do
\# geom_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))
#qqplot() +
# geom_bar(data=SumStatwtp, aes(phase, mean_Sleep,fill=Genotype), stat="identity", position=position_d
\# 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_dodge(width=0.5)
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>
#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
sleep_day <- wilcox.test(mean_sleep ~ Genotype, data=t_wt_days)</pre>
sleep_day
##
## Wilcoxon rank sum test with continuity correction
## data: mean_sleep by Genotype
## W = 27620, p-value = 0.2265
\#\# 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)$contrasts
## phase = dark:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT -0.0145 0.0218 56.7 -0.666 0.5080
##
## phase = light:
                          SE df t.ratio p.value
## contrast estimate
## McKO - WT -0.0443 0.0218 56.7 -2.038 0.0463
## Degrees-of-freedom method: kenward-roger
#emmeans::emmeans(lmer_wake, pairwise ~ phase | Genotype)
lmer_sleep <- lmer(mean_sleep ~ phase*Genotype+(1|Transmitter), data = t_wt_days)</pre>
#plot(lmer_sleep, type=c("p", "smooth"), col.line=1)
#lattice::ggmath(lmer sleep)
anova(lmer_sleep)
## Analysis of Variance Table
##
                 npar Sum Sq Mean Sq F value
## phase
                    1 10.8130 10.8130 432.6017
                    1 0.0937 0.0937
## Genotype
                                        3.7486
## phase:Genotype
                    1 0.0243 0.0243
                                       0.9719
emmeans::emmeans(lmer_sleep, pairwise ~ Genotype | phase)$contrasts
## phase = dark:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT 0.0161 0.0217 57.9 0.741 0.4614
##
## phase = light:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT 0.0453 0.0217 57.9
                                    2.088 0.0412
##
```

```
## Degrees-of-freedom method: kenward-roger
#emmeans::emmeans(lmer_sleep, pairwise ~ phase | Genotype)
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
                    1 8.6962 8.6962 405.4033
## phase
## Genotype
                    1 0.0442 0.0442 2.0607
## phase:Genotype
                    1 0.0208 0.0208 0.9675
emmeans::emmeans(lmer_sws, pairwise ~ Genotype | phase)$contrasts
## phase = dark:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT 0.0218 0.0282 29.1
                                   0.774 0.4451
##
## phase = light:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT 0.0488 0.0282 29.1
                                   1.733 0.0937
##
## Degrees-of-freedom method: kenward-roger
#emmeans::emmeans(lmer_sws, pairwise ~ phase | Genotype)
lmer_ps <- lmer(mean_ps ~ phase*Genotype+(1|Transmitter), data = t_wt_days)</pre>
#plot(lmer_ps, type=c("p","smooth"), col.line=1)
#lattice::qqmath(lmer_ps)
anova(lmer_ps)
## Analysis of Variance Table
                 npar
##
                        Sum Sq Mean Sq F value
                    1 0.115184 0.115184 178.5871
## phase
                    1 0.000092 0.000092 0.1430
## Genotype
                    1 0.000139 0.000139
## phase:Genotype
                                          0.2159
emmeans::emmeans(lmer_ps, pairwise ~ Genotype | phase)$contrasts
## phase = dark:
## contrast estimate
                          SE df t.ratio p.value
## McKO - WT -0.00572 0.0124 18.3 -0.460 0.6508
## phase = light:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT -0.00350 0.0124 18.3 -0.282 0.7811
## 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
                23 13.4075 0.58293 32.0622
## ZT
## Genotype
                 1 0.0615 0.06153 3.3843
## 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:
                          SE df t.ratio p.value
## contrast
             estimate
## McKO - WT -1.28e-01 0.0627 403 -2.040 0.0420
##
## ZT = 8:
                          SE df t.ratio p.value
## contrast
              estimate
## 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:
## contrast estimate
                       SE df t.ratio p.value
## 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:
## contrast estimate
                       SE df t.ratio p.value
## McKO - WT -2.34e-01 0.0627 403 -3.725 0.0002
##
## ZT = 15:
## contrast estimate SE df t.ratio p.value
## McKO - WT -1.65e-01 0.0627 403 -2.632 0.0088
## ZT = 16:
                      SE df t.ratio p.value
## contrast estimate
## 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:
## contrast estimate
                         SE df t.ratio p.value
## 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_sleep_zt <- lmer(mean_sleep ~ ZT*Genotype+(1|Transmitter), data = t_wt_days)</pre>
#plot(lmer_sleep_zt, type=c("p", "smooth"), col.line=1)
#lattice::qqmath(lmer_sleep_zt)
anova(lmer_sleep_zt)
## Analysis of Variance Table
##
              npar Sum Sq Mean Sq F value
## ZT
                23 13.5438 0.58886 32.0848
## Genotype
                 1 0.0688 0.06880 3.7486
## ZT:Genotype
                23 0.9905 0.04306 2.3464
#emmeans::emmeans(lmer_sleep_zt, pairwise ~ ZT / Genotype)
emmeans::emmeans(lmer_sleep_zt, pairwise ~ Genotype | ZT)$contrasts
## ZT = 0:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT 0.12347 0.063 403 1.961 0.0506
##
## ZT = 1:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT 0.00114 0.063 403
                                 0.018 0.9856
##
## ZT = 2:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT 0.06832 0.063 403
                                1.085 0.2786
##
## ZT = 3:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT 0.04304 0.063 403 0.683 0.4947
##
## ZT = 4:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT 0.10262 0.063 403
                                1.630 0.1039
##
## ZT = 5:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT -0.00923 0.063 403 -0.147 0.8835
##
## ZT = 6:
                         SE df t.ratio p.value
## contrast estimate
## McKO - WT -0.00157 0.063 403 -0.025 0.9801
##
## ZT = 7:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT 0.12970 0.063 403 2.060 0.0401
## ZT = 8:
## contrast estimate
                         SE df t.ratio p.value
## McKO - WT 0.05503 0.063 403
                                0.874 0.3826
##
## ZT = 9:
## contrast estimate
                         SE df t.ratio p.value
```

McKO - WT 0.03121 0.063 403 0.496 0.6204

```
##
## ZT = 10:
## contrast estimate
                      SE df t.ratio p.value
## McKO - WT -0.00873 0.063 403 -0.139 0.8899
## ZT = 11:
                      SE df t.ratio p.value
## contrast estimate
## McKO - WT 0.00906 0.063 403 0.144 0.8857
##
## ZT = 12:
## contrast estimate SE df t.ratio p.value
## McKO - WT -0.11864 0.063 403 -1.884 0.0603
## ZT = 13:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.09128 0.063 403 1.450 0.1479
##
## ZT = 14:
## contrast estimate
                      SE df t.ratio p.value
## McKO - WT 0.23570 0.063 403 3.743 0.0002
##
## ZT = 15:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.16700 0.063 403 2.652 0.0083
##
## ZT = 16:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.05756 0.063 403 0.914 0.3612
##
## ZT = 17:
## contrast estimate
                      SE df t.ratio p.value
## McKO - WT -0.06056 0.063 403 -0.962 0.3367
##
## ZT = 18:
## contrast estimate
                      SE df t.ratio p.value
## McKO - WT -0.17039 0.063 403 -2.706 0.0071
##
## ZT = 19:
## contrast estimate
                      SE df t.ratio p.value
## McKO - WT -0.09334 0.063 403 -1.482 0.1390
##
## ZT = 20:
## contrast estimate
                      SE df t.ratio p.value
## McKO - WT -0.07738 0.063 403 -1.229 0.2199
## ZT = 21:
## contrast estimate
                        SE df t.ratio p.value
## McKO - WT 0.01241 0.063 403 0.197 0.8439
##
## ZT = 22:
## contrast estimate SE df t.ratio p.value
## McKO - WT 0.00244 0.063 403 0.039 0.9691
##
## ZT = 23:
```

```
## contrast estimate
                       SE df t.ratio p.value
## McKO - WT 0.14716 0.063 403 2.337 0.0199
##
## Degrees-of-freedom method: kenward-roger
lmer_sws_zt <- lmer(mean_sws ~ ZT*Genotype+(1|Transmitter), data = t_wt_days)</pre>
{\it \#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
## ZT:Genotype 23 0.8769 0.03813 2.4418
#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:
## contrast estimate SE df t.ratio p.value
## 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:
                         SE df t.ratio p.value
## contrast estimate
## 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:
                         SE df t.ratio p.value
## contrast estimate
## 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:
                        SE df t.ratio p.value
## contrast estimate
## 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:
## contrast estimate SE df t.ratio p.value
## 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:
                         SE df t.ratio p.value
## contrast estimate
## 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:
## contrast estimate
                          SE
                              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:
```

```
## contrast estimate SE 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
                         SE
             estimate
                              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
```