Analysis of B-endorphin/ChAT IHC Count Data

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```
library(knitr)
library(tidyverse)
library(dplyr)
library(WebPower)
library(ggpubr)
library(car)
library(ggthemes)
library(viridis)
library(writexl)
library(DescTools)
```

Reading in and Tidying the Data

```
counts <- read_csv("../data/B-endo-totals.csv")</pre>
## Rows: 12 Columns: 13
## -- Column specification -------
## Delimiter: ","
## chr (3): Light, Time, Sex
## dbl (10): Mouse, ChAT, Bendo, Percentage, Age, GCLchat, GCLbendo, INLchat, I...
##
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
counts <- counts %>%
 dplyr::select(-ChAT, -Bendo, -Percentage) %>%
 #make columns lowercase
 rename(mouse = Mouse,
       light = Light,
       time = Time,
       sex = Sex,
       age = Age,
       GCL_chat = GCLchat,
       GCL_bendo = GCLbendo,
       INL_chat = INLchat,
```

```
INL_bendo = INLbendo) %>%
  #update the classes of some columns
  mutate(mouse = as.factor(mouse),
         light = factor(as.factor(light), levels=c("light", "dark")),
         time = as.factor(time),
  #reintroduce total counts
         total_chat = GCL_chat+INL_chat,
         total bendo = GCL bendo+INL bendo) %>%
  #calculate percentages
  mutate(GCL_perc=GCL_bendo/GCL_chat*100,
         INL_perc=INL_bendo/INL_chat*100,
         total_perc=total_bendo/total_chat*100)
counts <- counts %>% mutate(
  GCL_chat_mm = GCL_chat/Area,
  GCL_bendo_mm = GCL_bendo/Area,
  INL_chat_mm = INL_chat/Area,
  INL_bendo_mm = INL_bendo/Area
)
counts <- counts \%\% mutate(group = as.factor(c(1,1,1,4,4,4,2,2,2,3,3,3)))
counts
## # A tibble: 12 x 20
##
     mouse light time sex
                                age GCL_chat GCL_bendo INL_chat INL_bendo Area
##
      <fct> <fct> <fct> <chr> <dbl>
                                       <dbl>
                                                 <dbl>
                                                          <dbl>
                                                                    <dbl> <dbl>
## 1 11
                       F
                                 95
                                        4427
                                                   139
                                                           5589
                                                                      416 5.22
           light day
## 2 5
           light day
                                141
                                        1646
                                                   124
                                                           2013
                                                                      157 2.44
                       Μ
## 3 32
                                        2722
                                                   150
                                                           3232
                                                                      194 3.34
           light day
                      M
                                 65
                                                                      394 3.31
## 4 57
           dark night F
                                 54
                                        2474
                                                   173
                                                           3083
## 5 47
           dark night F
                                 97
                                        4374
                                                   385
                                                           5134
                                                                     1104 6.32
## 6 20
           dark night F
                                122
                                        5088
                                                   272
                                                           5886
                                                                     1075 5.64
                                                                     1575 4.43
## 7 31
                                 70
                                                   537
           dark day
                                        4515
                                                           4730
                       Μ
## 8 34
           dark day
                        F
                                 70
                                        4113
                                                   505
                                                           4587
                                                                     1005 4.54
## 9 97
            dark day
                       Μ
                                 74
                                        4373
                                                   462
                                                           5313
                                                                      836 5.62
## 10 92
            light night F
                                 96
                                        4221
                                                   501
                                                           5266
                                                                     1347 5.17
## 11 44
                                 41
                                        4986
                                                   465
                                                           5242
                                                                     1087 3.56
            light night M
## 12 93
            light night F
                                 96
                                        3138
                                                   225
                                                           3672
                                                                       348 6.16
## # ... with 10 more variables: total_chat <dbl>, total_bendo <dbl>,
      GCL_perc <dbl>, INL_perc <dbl>, total_perc <dbl>, GCL_chat_mm <dbl>,
## #
       GCL_bendo_mm <dbl>, INL_chat_mm <dbl>, INL_bendo_mm <dbl>, group <fct>
levels(counts$group)
## [1] "1" "2" "3" "4"
gcl_group_lm <- lm(GCL_bendo_mm ~ group, data=counts)</pre>
ScheffeTest(aov(gcl_group_lm), contrasts=c(1, -1/3, -1/3, -1/3))
##
##
    Posthoc multiple comparisons of means: Scheffe Test
##
       95% family-wise confidence level
##
## $group
```

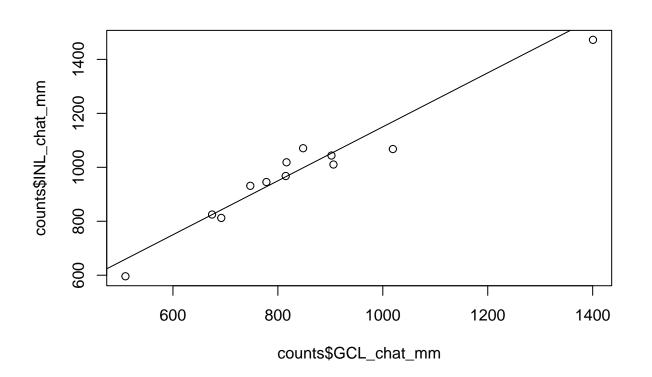
```
diff
                       lwr.ci
                                upr.ci pval
## 1-2,3,4 -41.44933 -103.9752 21.07657 0.2274
##
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
ScheffeTest(aov(gcl_group_lm), contrasts=c(1, 0, -1/2, -1/2))
##
##
    Posthoc multiple comparisons of means: Scheffe Test
##
      95% family-wise confidence level
##
## $group
##
              diff
                     lwr.ci
                               upr.ci
                                        pval
## 1-3,4 -30.12388 -96.44261 36.19485 0.5096
##
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
inl_group_lm <- lm(INL_bendo_mm ~ group, data=counts)</pre>
ScheffeTest(aov(inl_group_lm), contrasts=c(1, -1/3, -1/3, -1/3))
##
##
     Posthoc multiple comparisons of means: Scheffe Test
##
       95% family-wise confidence level
##
## $group
##
                diff
                        lwr.ci
                                 upr.ci
                                          pval
## 1-2,3,4 -136.2197 -338.3199 65.88044 0.2168
##
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
ScheffeTest(aov(inl_group_lm), contrasts=c(1, 0, -1/2, -1/2))
##
##
     Posthoc multiple comparisons of means: Scheffe Test
##
       95% family-wise confidence level
##
## $group
              diff
                     lwr.ci
                               upr.ci
## 1-3,4 -117.0748 -331.4344 97.28483 0.3659
##
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
#manova for ChAT cells per mm2
counts %>% group_by(light, time) %>% summarise(
 min_INL_chat_mm = min(INL_chat_mm),
 mean_INL_chat_mm = mean(INL_chat_mm),
 max_INL_chat_mm = max(INL_chat_mm),
```

```
min_GCL_chat_mm = min(GCL_chat_mm),
 mean_GCL_chat_mm = mean(GCL_chat_mm),
 max_GCL_chat_mm = max(GCL_chat_mm)
)
## 'summarise()' has grouped output by 'light'. You can override using the '.groups' argument.
## # A tibble: 4 x 8
## # Groups: light [2]
   light time min_INL_chat_mm mean_INL_chat_mm max_INL_chat_mm min_GCL_chat_mm
    <fct> <fct>
                           <dbl>
                                            <dbl>
                                                            <dbl>
## 1 light day
                            825
                                             954.
                                                            1071.
                                                                              675.
## 2 light night
                            596.
                                            1029.
                                                            1472.
                                                                              509.
                                            1008.
                                                            1068.
## 3 dark day
                            945.
                                                                              778.
## 4 dark night
                            812.
                                             929.
                                                            1044.
                                                                              692.
## # ... with 2 more variables: mean_GCL_chat_mm <dbl>, max_GCL_chat_mm <dbl>
counts %>% summarise(
 mean_INL_chat_mm = mean(INL_chat_mm),
 sd_INL_chat_mm = sd(INL_chat_mm),
 mean_GCL_chat_mm = mean(GCL_chat_mm),
 sd_GCL_chat_mm = sd(GCL_chat_mm)
)
## # A tibble: 1 x 4
   mean_INL_chat_mm sd_INL_chat_mm mean_GCL_chat_mm sd_GCL_chat_mm
##
                <dbl>
                              <dbl>
                                                <dbl>
                                                               <dbl>
## 1
                 980.
                                206.
                                                 842.
                                                                219.
shapiro.test(counts$GCL_chat_mm)
##
## Shapiro-Wilk normality test
##
## data: counts$GCL_chat_mm
## W = 0.88757, p-value = 0.1096
shapiro.test(counts$INL_chat_mm)
##
## Shapiro-Wilk normality test
## data: counts$INL_chat_mm
## W = 0.89834, p-value = 0.151
counts_long <- counts %>%
  #give one row for each GCL, INL, and total
  #retain only the percentages, not the raw counts
 dplyr::select(-c(GCL chat:INL bendo, total chat:total perc)) %>%
 pivot_longer(GCL_chat_mm:INL_bendo_mm,
```

```
names_to="type",
               values_to="cells_mm") %>%
  separate(type, sep="_", into=c("layer", "protein", "size")) %>%
  dplyr::select(-size) %>%
  mutate(layer = as.factor(layer),
         protein = as.factor(protein))
#testing manova assumptions
#multivariate normality
#normal distribution for each combo of indep and dependent variables
counts_long %>% filter(layer =="GCL" & protein == "chat" & time == "day") %>% pull(cells_mm) %>% shapir
##
##
   Shapiro-Wilk normality test
##
## data:
## W = 0.99202, p-value = 0.9935
counts_long %>% filter(layer =="GCL" & protein == "chat" & time == "night") %>% pull(cells_mm) %>% shap
##
## Shapiro-Wilk normality test
##
## data:
## W = 0.88836, p-value = 0.3097
counts_long %>% filter(layer =="INL" & protein == "chat" & time == "day") %>% pull(cells_mm) %>% shapir
## Shapiro-Wilk normality test
##
## data:
## W = 0.9109, p-value = 0.4424
counts_long %>% filter(layer =="INL" & protein == "chat" & time == "night") %>% pull(cells_mm) %>% shap
## Shapiro-Wilk normality test
##
## data:
## W = 0.95001, p-value = 0.7404
counts_long %>% filter(layer =="GCL" & protein == "chat" & light == "light") %>% pull(cells_mm) %>% sha
##
## Shapiro-Wilk normality test
##
## data: .
## W = 0.85537, p-value = 0.1739
```

```
counts_long %>% filter(layer =="GCL" & protein == "chat" & light == "dark") %>% pull(cells_mm) %>% shap
##
## Shapiro-Wilk normality test
## data: .
## W = 0.94687, p-value = 0.7149
counts_long %>% filter(layer =="INL" & protein == "chat" & light == "light") %>% pull(cells_mm) %>% sha
##
## Shapiro-Wilk normality test
##
## data:
## W = 0.95916, p-value = 0.8133
counts_long %>% filter(layer =="INL" & protein == "chat" & light == "dark") %>% pull(cells_mm) %>% shap
##
## Shapiro-Wilk normality test
##
## data: .
## W = 0.93201, p-value = 0.5957
#checking homogenous variance in each combo
car::leveneTest(counts$GCL_chat_mm, group=counts$light)
## Levene's Test for Homogeneity of Variance (center = median)
        Df F value Pr(>F)
## group 1 0.6201 0.4493
##
        10
car::leveneTest(counts$INL_chat_mm, group=counts$light)
## Levene's Test for Homogeneity of Variance (center = median)
       Df F value Pr(>F)
## group 1 2.2117 0.1678
##
        10
car::leveneTest(counts$GCL_chat_mm, group=counts$time)
## Levene's Test for Homogeneity of Variance (center = median)
        Df F value Pr(>F)
## group 1 1.3156 0.2781
##
        10
car::leveneTest(counts$INL_chat_mm, group=counts$time)
```

```
## Levene's Test for Homogeneity of Variance (center = median)
         Df F value Pr(>F)
## group 1 2.5293 0.1428
##
         10
#checking for multicollinearity
#low values are good
inllm <- lm(INL_chat_mm ~ light * time, counts)</pre>
car::vif(inllm)
##
        light
                    time light:time
##
gcllm <- lm(GCL_chat_mm ~ light * time, counts)</pre>
car::vif(gcllm)
##
        light
                    time light:time
##
            2
                        2
#linearity of dep variables
plot(counts$GCL_chat_mm, counts$INL_chat_mm)
abline(150,1)
```

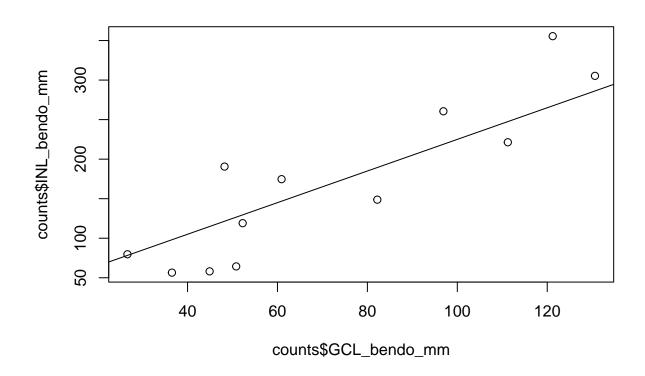


```
#good to go on manova!
mantest <- manova(cbind(GCL_chat_mm, INL_chat_mm) ~ light * time, data=counts)</pre>
summary(mantest)
                  Pillai approx F num Df den Df Pr(>F)
##
## light
              1 0.054947 0.20350
                                       2
                                             7 0.8205
                                       2
## time
              1 0.005750 0.02024
                                              7 0.9800
## light:time 1 0.254205 1.19298
                                       2
                                              7 0.3582
## Residuals
              8
summary.aov(mantest)
## Response GCL_chat_mm :
              Df Sum Sq Mean Sq F value Pr(>F)
##
                             31 0.0005 0.9825
## light
                     31
               1
## time
               1
                     62
                             62 0.0010 0.9752
              1 46922
                          46922 0.7826 0.4021
## light:time
## Residuals
               8 479633
                          59954
##
  Response INL_chat_mm :
##
              Df Sum Sq Mean Sq F value Pr(>F)
## light
                   1626
                           1626 0.0290 0.8691
## time
               1
                     13
                             13 0.0002 0.9884
## light:time
              1 17622
                          17622 0.3141 0.5905
## Residuals
               8 448865
                          56108
#light:time interaction is significant overall and for GCL and INL
#look at comparisons through univariate models for each dep variable
anova(gcllm)
## Analysis of Variance Table
##
## Response: GCL_chat_mm
             Df Sum Sq Mean Sq F value Pr(>F)
##
## light
                 31 31 0.0005 0.9825
                            62 0.0010 0.9752
## time
              1
                    62
## light:time 1 46922
                        46922 0.7826 0.4021
## Residuals
              8 479633
                         59954
anova(inllm)
## Analysis of Variance Table
## Response: INL_chat_mm
##
             Df Sum Sq Mean Sq F value Pr(>F)
             1 1626 1626 0.0290 0.8691
## light
## time
              1
                    13
                          13 0.0002 0.9884
## light:time 1 17622
                         17622 0.3141 0.5905
## Residuals
              8 448865
                         56108
```

```
#manova for bendo cells per mm2
counts_sumstats <- counts %>% group_by(time, light) %>%
  summarise(mean_INL_bendo_mm = mean(INL_bendo_mm),
 sd_INL_bendo_mm = sd(INL_bendo_mm),
 mean_GCL_bendo_mm = mean(GCL_bendo_mm),
 sd_GCL_bendo_mm = sd(GCL_bendo_mm)
## 'summarise()' has grouped output by 'time'. You can override using the '.groups' argument.
shapiro.test(counts$GCL_bendo_mm)
##
## Shapiro-Wilk normality test
## data: counts$GCL_bendo_mm
## W = 0.90829, p-value = 0.2028
shapiro.test(counts$INL_bendo_mm)
##
   Shapiro-Wilk normality test
## data: counts$INL bendo mm
## W = 0.92663, p-value = 0.3458
#testing manova assumptions
#multivariate normality
#normal distribution for each combo of indep and dependent variables
counts_long %>% filter(layer =="GCL" & protein == "bendo" & time == "day") %>% pull(cells_mm) %>% shapi
##
## Shapiro-Wilk normality test
##
## data:
## W = 0.92539, p-value = 0.545
counts_long %>% filter(layer =="GCL" & protein == "bendo" & time == "night") %>% pull(cells_mm) %>% sha
## Shapiro-Wilk normality test
##
## data:
## W = 0.87924, p-value = 0.2656
counts_long %>% filter(layer =="INL" & protein == "bendo" & time == "day") %>% pull(cells_mm) %>% shapi.
##
## Shapiro-Wilk normality test
##
## data: .
## W = 0.85792, p-value = 0.1821
```

```
counts_long %>% filter(layer =="INL" & protein == "bendo" & time == "night") %>% pull(cells_mm) %>% sha
##
## Shapiro-Wilk normality test
##
## data: .
## W = 0.98211, p-value = 0.9616
counts_long %>% filter(layer =="GCL" & protein == "bendo" & light == "light") %>% pull(cells_mm) %>% sh
##
## Shapiro-Wilk normality test
##
## data: .
## W = 0.86634, p-value = 0.212
counts_long %>% filter(layer =="GCL" & protein == "bendo" & light == "dark") %>% pull(cells_mm) %>% sha
##
## Shapiro-Wilk normality test
##
## data: .
## W = 0.88645, p-value = 0.3
counts_long %>% filter(layer =="INL" & protein == "bendo" & light == "dark") %>% pull(cells_mm) %>% sha
##
## Shapiro-Wilk normality test
##
## data: .
## W = 0.87501, p-value = 0.2469
#did not pass, but transformations don't either
counts_long %>% filter(layer =="INL" & protein == "bendo" & light == "light") %>% pull(cells_mm) %>% sh
## Shapiro-Wilk normality test
##
## data: .
## W = 0.73385, p-value = 0.01378
#checking homogenous variance in each combo
car::leveneTest(counts$GCL_bendo_mm, group=counts$light)
## Levene's Test for Homogeneity of Variance (center = median)
        Df F value Pr(>F)
## group 1 0.0384 0.8485
##
         10
```

```
car::leveneTest(counts$INL_bendo_mm, group=counts$light)
## Levene's Test for Homogeneity of Variance (center = median)
        Df F value Pr(>F)
## group 1 0.2279 0.6434
##
        10
car::leveneTest(counts$GCL_bendo_mm, group=counts$time)
## Levene's Test for Homogeneity of Variance (center = median)
       Df F value Pr(>F)
## group 1 0.2692 0.6152
##
        10
car::leveneTest(counts$INL_bendo_mm, group=counts$time)
## Levene's Test for Homogeneity of Variance (center = median)
        Df F value Pr(>F)
## group 1 0.2504 0.6276
##
        10
#checking for multicollinearity
#low values are good
inl_bendo_lm <- lm(INL_bendo_mm ~ light * time, counts)</pre>
car::vif(inl_bendo_lm)
##
       light
                  time light:time
##
           2
                       2
gcl_bendo_lm <- lm(GCL_bendo_mm ~ light * time, counts)</pre>
car::vif(gcl_bendo_lm)
##
       light
                 time light:time
##
           2
                      2
#linearity of dep variables
plot(counts$GCL_bendo_mm, counts$INL_bendo_mm)
abline(25,2)
```



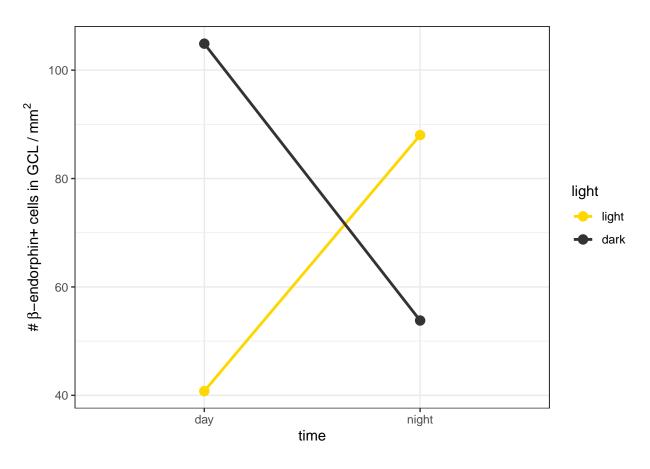
```
#good to go on manova!
#manova
man_bendo_test <- manova(cbind(GCL_bendo_mm, INL_bendo_mm) ~ light * time, data=counts)</pre>
summary(man_bendo_test)
              Df Pillai approx F num Df den Df Pr(>F)
##
                           0.7502
## light
               1 0.17650
                                       2
                                               7 0.50678
## time
               1 0.19087
                           0.8256
                                        2
                                               7 0.47651
                                        2
                                               7 0.04767 *
## light:time
              1 0.58085
                           4.8503
## Residuals
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
summary.aov(man_bendo_test)
##
   {\tt Response~GCL\_bendo\_mm}~:
##
               Df Sum Sq Mean Sq F value Pr(>F)
## light
                1 669.9
                           669.9 0.9291 0.36334
## time
                    11.1
                            11.1 0.0154 0.90419
## light:time
                          7249.1 10.0529 0.01318 *
                1 7249.1
## Residuals
                8 5768.8
                           721.1
## Signif. codes:
                   0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
```

```
## Response INL_bendo_mm :
##
              Df Sum Sq Mean Sq F value Pr(>F)
                         12383 1.6436 0.23573
## light
              1 12383
                          2668 0.3541 0.56825
                   2668
## time
               1
## light:time
               1 36474
                          36474 4.8415 0.05896 .
## Residuals
               8 60270
                           7534
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
#light:time interaction is significant overall and for GCL, close in INL
#look at comparisons through univariate models for each dep variable
anova(gcl_bendo_lm)
## Analysis of Variance Table
## Response: GCL bendo mm
             Df Sum Sq Mean Sq F value Pr(>F)
##
              1 669.9
                       669.9 0.9291 0.36334
## light
                  11.1
## time
                         11.1 0.0154 0.90419
              1
## light:time 1 7249.1 7249.1 10.0529 0.01318 *
              8 5768.8
## Residuals
                       721.1
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
#summary(gcl_bendo_lm)
Anova(gcl_bendo_lm, type="3")
## Anova Table (Type III tests)
## Response: GCL_bendo_mm
              Sum Sq Df F value Pr(>F)
## (Intercept) 4990.5 1 6.9207 0.03014 *
## light
              6163.3 1 8.5470 0.01919 *
              3346.1 1 4.6402 0.06337 .
## time
## light:time 7249.1 1 10.0529 0.01318 *
## Residuals 5768.8 8
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
anova(inl_bendo_lm)
## Analysis of Variance Table
## Response: INL_bendo_mm
##
             Df Sum Sq Mean Sq F value Pr(>F)
              1 12383
                       12383 1.6436 0.23573
## light
## time
              1
                 2668
                          2668 0.3541 0.56825
## light:time 1 36474
                         36474 4.8415 0.05896 .
## Residuals
              8 60270
                         7534
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

```
Anova(inl_bendo_lm, type="3")
## Anova Table (Type III tests)
##
## Response: INL_bendo_mm
##
              Sum Sq Df F value Pr(>F)
## (Intercept) 13618 1 1.8076 0.21567
## light
               45680 1 6.0635 0.03917 *
               29435 1 3.9071 0.08348 .
## time
## light:time 36474 1 4.8415 0.05896 .
## Residuals
               60270 8
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
summary(inl_bendo_lm)
##
## Call:
## lm(formula = INL_bendo_mm ~ light * time, data = counts)
## Residuals:
##
       Min
                 1Q
                      Median
                                   3Q
## -150.964 -25.990
                       4.645
                               35.143 113.647
## Coefficients:
##
                      Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                         67.37
                                   50.11
                                           1.344
                                                  0.2157
                                           2.462
## lightdark
                        174.51
                                   70.87
                                                   0.0392 *
## timenight
                        140.08
                                   70.87
                                            1.977
                                                   0.0835 .
## lightdark:timenight -220.53
                                   100.22 -2.200
                                                   0.0590 .
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 86.8 on 8 degrees of freedom
## Multiple R-squared: 0.4609, Adjusted R-squared: 0.2587
## F-statistic: 2.28 on 3 and 8 DF, p-value: 0.1563
emmeans::emmeans(gcl_bendo_lm, pairwise ~ light | time)
## $emmeans
## time = day:
## light emmean
                 SE df lower.CL upper.CL
          40.8 15.5 8
                           5.03
                                    76.5
## light
## dark
         104.9 15.5 8
                           69.13
                                    140.6
##
## time = night:
## light emmean
                  SE df lower.CL upper.CL
           88.0 15.5 8
## light
                           52.26
                                  123.8
                                    89.6
## dark
           53.8 15.5 8
                           18.05
##
## Confidence level used: 0.95
```

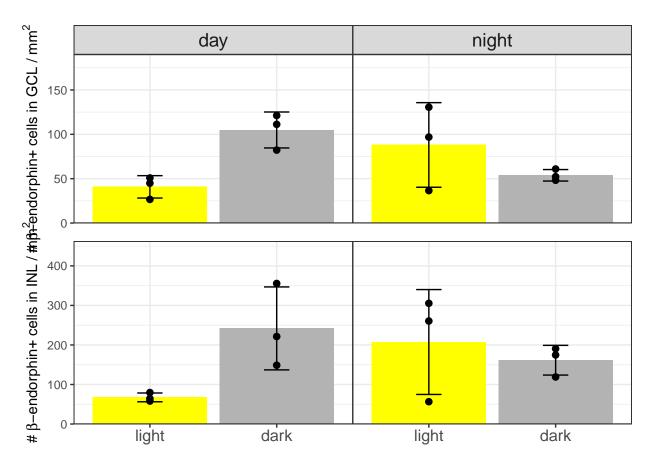
```
##
## $contrasts
## time = day:
## contrast
              estimate SE df t.ratio p.value
## light - dark -64.1 21.9 8 -2.924 0.0192
##
## time = night:
## contrast
               estimate SE df t.ratio p.value
## light - dark
                   34.2 21.9 8
                                1.560 0.1573
emmeans::emmeans(gcl_bendo_lm, pairwise ~ time | light)
## $emmeans
## light = light:
## time emmean
                SE df lower.CL upper.CL
          40.8 15.5 8
                       5.03
## night 88.0 15.5 8
                         52.26
                                  123.8
##
## light = dark:
## time emmean
                SE df lower.CL upper.CL
        104.9 15.5 8
                                140.6
## day
                         69.13
## night 53.8 15.5 8
                         18.05
                                  89.6
##
## Confidence level used: 0.95
##
## $contrasts
## light = light:
## contrast estimate SE df t.ratio p.value
## day - night -47.2 21.9 8 -2.154 0.0634
##
## light = dark:
## contrast
            estimate SE df t.ratio p.value
## day - night 51.1 21.9 8 2.330 0.0482
emmeans::emmeans(inl_bendo_lm, pairwise ~ light | time)
## $emmeans
## time = day:
## light emmean SE df lower.CL upper.CL
## light 67.4 50.1 8 -48.2
## dark 241.9 50.1 8
                         126.3
                                    357
##
## time = night:
## light emmean
                 SE df lower.CL upper.CL
## light 207.5 50.1 8
                          91.9
                                    323
## dark 161.4 50.1 8
                          45.9
                                    277
##
## Confidence level used: 0.95
##
## $contrasts
## time = day:
## contrast
              estimate SE df t.ratio p.value
                 -175 70.9 8 -2.462 0.0392
## light - dark
```

```
##
## time = night:
## contrast
                estimate SE df t.ratio p.value
## light - dark
                   46 70.9 8 0.649 0.5343
emmeans::emmeans(inl_bendo_lm, pairwise ~ time | light)
## $emmeans
## light = light:
## time emmean
                 SE df lower.CL upper.CL
## day
           67.4 50.1 8
                           -48.2
                                      183
## night 207.5 50.1 8
                            91.9
                                      323
##
## light = dark:
## time emmean
                 SE df lower.CL upper.CL
## day
          241.9 50.1 8
                           126.3
## night 161.4 50.1 8
                            45.9
                                      277
## Confidence level used: 0.95
##
## $contrasts
## light = light:
## contrast
              estimate SE df t.ratio p.value
## day - night -140.1 70.9 8 -1.977 0.0835
##
## light = dark:
## contrast
             estimate SE df t.ratio p.value
## day - night
                   80.4 70.9 8 1.135 0.2892
gcl_inter_plot <- ggplot() +</pre>
  stat_summary(data=counts, aes(x = time, y = GCL_bendo_mm, color = light, group = light),
              fun = mean, geom = "point", size=3) +
  stat_summary(data=counts, aes(x = time, y = GCL_bendo_mm, color = light, group = light),
              fun = mean, geom = "line", size=1) +
  #qeom_errorbar(data=counts_sumstats, aes(x=time, ymin=mean_GCL_bendo_mm-sd_GCL_bendo_mm,
                ymax=mean\_GCL\_bendo\_mm+sd\_GCL\_bendo\_mm), width=0.1) +
  labs(y=expression(paste("# ", beta, "-endorphin+ cells in GCL /" ~mm^2))) +
  scale_color_manual(values=c("gold", "grey20")) +
  theme bw()
gcl_inter_plot
```



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

```
gcl_mm_plot <- ggplot() +</pre>
  geom_bar(data=sumstats, aes(x=light, y=mean_gcl, fill=light), stat="identity") +
  geom_errorbar(data=sumstats, aes(x=light,
                               ymin=mean_gcl-sd_gcl,
                               ymax=mean_gcl+sd_gcl), width=0.2) +
  geom_point(data=counts, aes(x=light, y=GCL_bendo_mm), size=2) +
  facet_wrap(~time) +
  theme bw() %+replace%
  theme(axis.title.x=element blank(),
        axis.ticks.x = element_blank(),
        axis.text.x = element_blank(),
        panel.spacing = unit(0, "pt"),
        strip.text.x = element_text(size = 14),
        legend.position="none",
        plot.background = element_rect(fill = "transparent",colour = NA),
        panel.background = element_rect(fill = "transparent",colour = NA)) +
 labs(y=expression(paste("# ", beta, "-endorphin+ cells in GCL /" ~mm^2))) +
scale_fill_manual(values=c("yellow", "gray70")) +
  scale_y = continuous(expand = expansion(mult = c(0, 0.4)))
inl_mm_plot <- ggplot() +</pre>
  geom_bar(data=sumstats, aes(x=light, y=mean_inl, fill=light), stat="identity") +
  geom_errorbar(data=sumstats, aes(x=light,
                               ymin=mean_inl-sd_inl,
                               ymax=mean_inl+sd_inl), width=0.2) +
  geom_point(data=counts, aes(x=light, y=INL_bendo_mm), size=2) +
  facet wrap(~time) +
  theme_bw() %+replace%
  theme(axis.title.x=element_blank(),
        axis.text.x = element_text(size = 12),
        panel.spacing = unit(0, "pt"),
        strip.background = element_blank(),
        strip.text.x = element_blank(),
        legend.position="none",
        plot.background = element_rect(fill = "transparent",colour = NA),
        panel.background = element_rect(fill = "transparent",colour = NA)) +
  labs(y=expression(paste("# ", beta, "-endorphin+ cells in INL /" ~mm^2))) +
  scale_fill_manual(values=c("yellow", "gray70")) +
  scale_y\_continuous(expand = expansion(mult = c(0, 0.3)))
both_mm_plots <- ggarrange(gcl_mm_plot, inl_mm_plot, nrow=2)</pre>
both_mm_plots
```



```
ggsave("../figures/both_mm_plots.png", plot=both_mm_plots, height=6, width=6, bg = "transparent")

ctrl_counts <- counts %>% filter(light == "light" & time == "day")

#between var

sum(3*(mean(counts$GCL_bendo_mm - mean(ctrl_counts$GCL_bendo_mm)))^2)
```

[1] 2899.204

```
#within.var
anova(gcl_bendo_lm)["Residuals", "Mean Sq"]
```

[1] 721.0983

```
power.anova.test(groups=4, n=3, between.var=2899.204,
within.var=721.0983, sig.level=0.05)
```

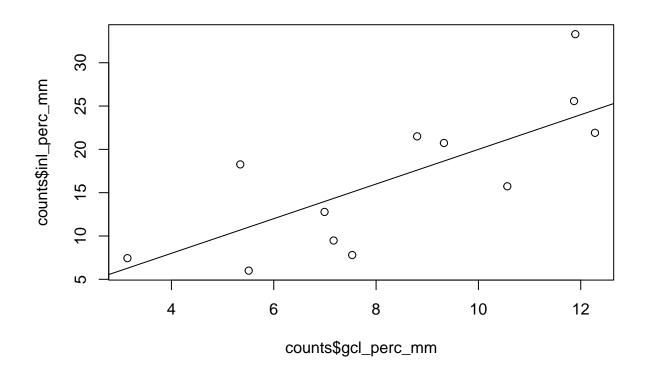
```
##
## Balanced one-way analysis of variance power calculation
##
## groups = 4
## n = 3
## between.var = 2899.204
## within.var = 721.0983
```

```
##
         sig.level = 0.05
##
             power = 0.9820665
##
## NOTE: n is number in each group
#between var
sum(3*(mean(counts$INL_bendo_mm - mean(ctrl_counts$INL_bendo_mm)))^2)
## [1] 31312.94
#within.var
anova(inl_bendo_lm)["Residuals", "Mean Sq"]
## [1] 7533.694
power.anova.test(groups=4, n=3, between.var=31312.94,
                 within.var=7533.694, sig.level=0.05)
##
##
        Balanced one-way analysis of variance power calculation
##
##
            groups = 4
                 n = 3
##
       between.var = 31312.94
##
##
        within.var = 7533.694
##
         sig.level = 0.05
             power = 0.9849714
##
## NOTE: n is number in each group
#percentage / mm2
counts <- counts %% mutate(inl_perc_mm = INL_bendo_mm/INL_chat_mm*100,</pre>
                            gcl_perc_mm = GCL_bendo_mm/GCL_chat_mm*100)
perc_sumstats <- counts %>% group_by(light, time) %>%
            summarise(n = n(),
                      mean_inl_perc = mean(inl_perc_mm),
                      sd_inl_perc = sd(inl_perc_mm),
                      mean_gcl_perc = mean(gcl_perc_mm),
                      sd_gcl_perc = sd(gcl_perc_mm))
## 'summarise()' has grouped output by 'light'. You can override using the '.groups' argument.
shapiro.test(counts$gcl_perc_mm)
##
## Shapiro-Wilk normality test
## data: counts$gcl_perc_mm
## W = 0.94973, p-value = 0.633
```

```
shapiro.test(counts$inl_perc_mm)
##
##
   Shapiro-Wilk normality test
##
## data: counts$inl_perc_mm
## W = 0.94583, p-value = 0.5771
\verb|counts_perc_long| <- \verb|counts| \%>\%
  #give one row for each GCL, INL, and total
  #retain only the percentages, not the raw counts
  dplyr::select(-c(GCL_chat:INL_bendo, total_chat:INL_bendo_mm)) %>%
  pivot_longer(inl_perc_mm:gcl_perc_mm,
               names_to="type",
               values_to="perc_mm") %>%
  separate(type, sep="_", into=c("layer", "na", "size")) %>%
  dplyr::select(-c(na,size)) %>%
 mutate(layer = as.factor(layer))
#testing manova assumptions
#multivariate normality
#normal distribution for each combo of indep and dependent variables
counts_perc_long %>% filter(layer =="gcl" & time == "day") %>% pull(perc_mm) %>% shapiro.test(.)
##
##
   Shapiro-Wilk normality test
## data:
## W = 0.9171, p-value = 0.4847
counts_perc_long %>% filter(layer =="gcl" & time == "night") %>% pull(perc_mm) %>% shapiro.test(.)
##
##
   Shapiro-Wilk normality test
## data:
## W = 0.96858, p-value = 0.8828
counts_perc_long %>% filter(layer =="gcl" & time == "day") %>% pull(perc_mm) %>% shapiro.test(.)
##
##
   Shapiro-Wilk normality test
## data:
## W = 0.9171, p-value = 0.4847
counts_perc_long %>% filter(layer =="gcl" & time == "night") %>% pull(perc_mm) %>% shapiro.test(.)
##
## Shapiro-Wilk normality test
```

```
##
## data:
## W = 0.96858, p-value = 0.8828
counts_perc_long %>% filter(layer =="gcl" & light == "light") %>% pull(perc_mm) %>% shapiro.test(.)
##
##
   Shapiro-Wilk normality test
##
## data: .
## W = 0.99193, p-value = 0.9934
counts_perc_long %>% filter(layer =="gcl" & light == "dark") %>% pull(perc_mm) %>% shapiro.test(.)
##
   Shapiro-Wilk normality test
##
##
## data:
## W = 0.93389, p-value = 0.6104
counts_perc_long %>% filter(layer =="inl" & light == "dark") %>% pull(perc_mm) %>% shapiro.test(.)
##
## Shapiro-Wilk normality test
##
## data: .
## W = 0.91084, p-value = 0.442
counts_perc_long %>% filter(layer =="inl" & light == "light") %>% pull(perc_mm) %>% shapiro.test(.)
##
## Shapiro-Wilk normality test
##
## data:
## W = 0.80407, p-value = 0.06392
#checking homogenous variance in each combo
car::leveneTest(counts$gcl_perc_mm, group=counts$light)
## Levene's Test for Homogeneity of Variance (center = median)
        Df F value Pr(>F)
## group 1 0.0151 0.9047
        10
car::leveneTest(counts$inl_perc_mm, group=counts$light)
## Levene's Test for Homogeneity of Variance (center = median)
        Df F value Pr(>F)
## group 1 0.0504 0.8268
        10
##
```

```
car::leveneTest(counts$gcl_perc_mm, group=counts$time)
## Levene's Test for Homogeneity of Variance (center = median)
        Df F value Pr(>F)
## group 1 2.5789 0.1394
##
         10
car::leveneTest(counts$inl_perc_mm, group=counts$time)
## Levene's Test for Homogeneity of Variance (center = median)
       Df F value Pr(>F)
## group 1 1.3768 0.2678
##
         10
#checking for multicollinearity
#low values are good
inl_perc_lm <- lm(inl_perc_mm ~ light * time, counts)</pre>
car::vif(inl_perc_lm)
                    time light:time
##
        light
##
                       2
gcl_perc_lm <- lm(gcl_perc_mm ~ light * time, counts)</pre>
car::vif(gcl_perc_lm)
##
        light
                    time light:time
##
#linearity of dep variables
plot(counts$gcl_perc_mm, counts$inl_perc_mm)
abline(0,2)
```



```
#good to go on manova!
#manova
man_perc_test <- manova(cbind(gcl_perc_mm, inl_perc_mm) ~ light * time, data=counts)</pre>
summary(man_perc_test)
##
              Df Pillai approx F num Df den Df Pr(>F)
                           2.0965
                                              7 0.19343
## light
               1 0.37461
                                       2
## time
               1 0.11699
                           0.4637
                                       2
                                              7 0.64697
                                       2
                                              7 0.02241 *
## light:time
              1 0.66218
                           6.8605
## Residuals
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
summary.aov(man_perc_test)
##
   Response gcl_perc_mm :
##
              Df Sum Sq Mean Sq F value
## light
                1 10.693 10.693 3.0196 0.120464
                           0.167 0.0471 0.833664
## time
                1 0.167
                1 55.374 55.374 15.6365 0.004211 **
## light:time
                8 28.331
                           3.541
## Residuals
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
```

```
## Response inl_perc_mm :
##
              Df Sum Sq Mean Sq F value Pr(>F)
## light
              1 179.82 179.818 4.2818 0.07231 .
               1 21.74 21.741 0.5177 0.49231
## time
## light:time
              1 233.58 233.584 5.5621 0.04607 *
## Residuals
               8 335.96 41.996
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
#light:time interaction is just about significant overall and for GCL and INL
#look at comparisons through univariate models for each dep variable
anova(gcl perc lm)
## Analysis of Variance Table
## Response: gcl_perc_mm
             Df Sum Sq Mean Sq F value
## light
              1 10.693 10.693 3.0196 0.120464
## time
              1 0.167
                        0.167 0.0471 0.833664
## light:time 1 55.374 55.374 15.6365 0.004211 **
## Residuals
              8 28.331
                         3.541
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
summary(gcl_perc_lm)
##
## lm(formula = gcl_perc_mm ~ light * time, data = counts)
##
## Residuals:
       Min
                 1Q
                     Median
                                   3Q
                                           Max
## -2.28500 -1.18579 0.03093 0.96322 2.41406
##
## Coefficients:
##
                      Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                         5.395
                                   1.086 4.965 0.00110 **
## lightdark
                         6.184
                                   1.537
                                           4.025 0.00382 **
## timenight
                         4.061
                                   1.537
                                           2.643 0.02959 *
## lightdark:timenight
                       -8.593
                                   2.173 -3.954 0.00421 **
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 1.882 on 8 degrees of freedom
## Multiple R-squared: 0.7004, Adjusted R-squared: 0.5881
## F-statistic: 6.234 on 3 and 8 DF, p-value: 0.01728
anova(inl_perc_lm)
## Analysis of Variance Table
## Response: inl_perc_mm
```

```
Df Sum Sq Mean Sq F value Pr(>F)
              1 179.82 179.818 4.2818 0.07231 .
## light
## time
              1 21.74 21.741 0.5177 0.49231
## light:time 1 233.58 233.584 5.5621 0.04607 *
## Residuals 8 335.96 41.996
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
summary(inl_perc_lm)
##
## Call:
## lm(formula = inl_perc_mm ~ light * time, data = counts)
## Residuals:
##
      Min
               1Q Median
                              3Q
## -9.1204 -2.4874 0.5396 2.6011 9.6505
## Coefficients:
                     Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                        7.082
                                   3.741 1.893 0.0950 .
## lightdark
                       16.566
                                   5.291
                                          3.131
                                                 0.0140 *
## timenight
                                   5.291
                                                 0.0612 .
                       11.516
                                          2.176
                                   7.483 - 2.358
## lightdark:timenight -17.648
                                                 0.0461 *
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 6.48 on 8 degrees of freedom
## Multiple R-squared: 0.5643, Adjusted R-squared: 0.4009
## F-statistic: 3.454 on 3 and 8 DF, p-value: 0.07147
emmeans::emmeans(gcl_perc_lm, pairwise ~ light * time, adjust="none")$contrasts
## contrast
                           estimate
                                     SE df t.ratio p.value
## light day - dark day
                              -6.18 1.54 8 -4.025 0.0038
## light day - light night
                              -4.06 1.54 8 -2.643 0.0296
## light day - dark night
                              -1.65 1.54 8 -1.075 0.3136
## dark day - light night
                               2.12 1.54 8
                                             1.382 0.2043
## dark day - dark night
                               4.53 1.54 8
                                             2.950 0.0184
## light night - dark night
                               2.41 1.54 8
                                             1.567 0.1557
emmeans::emmeans(inl_perc_lm, pairwise ~ light * time, adjust="none")$contrasts
## contrast
                                     SE df t.ratio p.value
                           estimate
                           -16.57 5.29 8 -3.131 0.0140
## light day - dark day
## light day - light night
                             -11.52 5.29 8 -2.176 0.0612
## light day - dark night
                             -10.43 5.29 8 -1.972 0.0841
## dark day - light night
                              5.05 5.29 8
                                            0.954 0.3678
## dark day - dark night
                               6.13 5.29 8
                                             1.159 0.2799
## light night - dark night 1.08 5.29 8 0.204 0.8431
```

```
gcl_perc_plot <- ggplot() +</pre>
  geom_bar(data=perc_sumstats, aes(x=light, y=mean_gcl_perc, fill=light), stat="identity") +
  geom_errorbar(data=perc_sumstats, aes(x=light,
                              ymin=mean_gcl_perc-sd_gcl_perc,
                              ymax=mean_gcl_perc+sd_gcl_perc), width=0.2) +
  geom_point(data=counts, aes(x=light, y=gcl_perc_mm)) +
  facet_wrap(~time) +
  theme bw() %+replace%
  theme(axis.title.x=element blank(),
        axis.ticks.x = element_blank(),
        axis.text.x = element_blank(),
        panel.spacing = unit(0, "pt"),
        legend.position="none") +
  labs(y=expression(paste("%", beta, "-endorphin+/ChAT+ cells in GCL /" ~mm^2))) +
  scale_fill_manual(values=c("yellow", "gray70")) +
  scale_y_continuous(expand = expansion(mult = c(0, 0.4)),
                     labels=function(x) paste0(x,"%"))
inl_perc_plot <- ggplot() +</pre>
  geom_bar(data=perc_sumstats, aes(x=light, y=mean_inl_perc, fill=light), stat="identity") +
  geom_errorbar(data=perc_sumstats, aes(x=light,
                              ymin=mean_inl_perc-sd_inl_perc,
                              ymax=mean_inl_perc+sd_inl_perc), width=0.2) +
  geom_point(data=counts, aes(x=light, y=inl_perc_mm)) +
  facet wrap(~time) +
  theme bw() %+replace%
  theme(axis.title.x=element_blank(),
        axis.ticks.x = element_blank(),
        panel.spacing = unit(0, "pt"),
        strip.background = element_blank(),
        strip.text.x = element_blank(),
        legend.position="none") +
  labs(y=expression(paste("%", beta, "-endorphin+/ChAT+ cells in INL /" ~mm^2))) +
  scale_fill_manual(values=c("yellow", "gray70")) +
  scale_y\_continuous(expand = expansion(mult = c(0, 0.3)),
                     labels=function(x) paste0(x,"%"))
#,
         plot.background = element_rect(fill = "transparent", colour = NA),
#
#
         panel.background = element_rect(fill = "transparent", colour = NA)
both_perc_plots <- ggarrange(gcl_perc_plot, inl_perc_plot, nrow=2)</pre>
ggsave("../figures/both_perc_plots.png", plot=both_perc_plots, height=7, width=6, bg = "transparent")
ctrl_counts <- counts %>% filter(light == "light" & time == "day")
#between var
sum(3*(mean(counts$gcl_perc_mm - mean(ctrl_counts$gcl_perc_mm)))^2)
```

[1] 26.5387

```
#within.var
anova(gcl_perc_lm)["Residuals", "Mean Sq"]
## [1] 3.541329
power.anova.test(groups=4, n=3, between.var=26.5387,
                 within.var=3.541329, sig.level=0.05)
##
##
        Balanced one-way analysis of variance power calculation
##
##
            groups = 4
##
                 n = 3
##
       between.var = 26.5387
##
        within.var = 3.541329
##
         sig.level = 0.05
##
             power = 0.9998598
##
## NOTE: n is number in each group
#between var
sum(3*(mean(counts$inl_perc_mm - mean(ctrl_counts$inl_perc_mm)))^2)
## [1] 278.1516
#within.var
anova(inl_perc_lm)["Residuals", "Mean Sq"]
## [1] 41.99559
power.anova.test(groups=4, n=3, between.var=278.1516,
                 within.var=41.99559, sig.level=0.05)
##
##
        Balanced one-way analysis of variance power calculation
##
##
            groups = 4
##
                 n = 3
##
       between.var = 278.1516
##
        within.var = 41.99559
##
         sig.level = 0.05
##
             power = 0.9995018
##
## NOTE: n is number in each group
summary_table <- counts %>% group_by(light, time) %>%
  summarise(n = n(),
            "Total # ChAT+ counted in INL" = sum(INL_chat),
            "Total # B-endo+ counted in INL" = sum(INL bendo),
            "Mean ChAT+ cells in INL/mm2" = pasteO(round(mean(INL_chat_mm), 0),
```

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

```
summary_table
```

```
## # A tibble: 4 x 11
## # Groups: light [2]
     light time
                    n 'Total # ChAT+ coun~ 'Total # B-endo+ c~ 'Mean ChAT+ cells ~
##
     <fct> <fct> <int>
                                      <dbl>
                                                          <dbl> <chr>
## 1 light day
                    3
                                      10834
                                                           767 954±123
## 2 light night
                     3
                                      14180
                                                           2782 1029±438
## 3 dark day
                     3
                                      14630
                                                           3416 1008±61
## 4 dark night
                    3
                                                           2573 929±116
                                      14103
## # ... with 5 more variables: Mean B-endo+ cells in INL/mm2 <chr>,
     Total # ChAT+ counted in GCL <dbl>, Total # B-endo+ counted in GCL <dbl>,
## #
      Mean ChAT+ cells in GCL/mm2 <chr>, Mean B-endo+ cells in GCL/mm2 <chr>
```

```
#write_xlsx(summary_table, "../data/bendo_summary_table.xlsx")
```

OLD

```
#data setup for previous anova analysis
#also used in manova shapiro tests for ease
anovpercs <- counts %>%
  #give one row for each GCL, INL, and total
  #retain only the percentages, not the raw counts
  dplyr::select(-c(GCL_chat:total_bendo)) %>%
  pivot_longer(GCL_perc:total_perc,
              names_to="type",
              values_to="perc") %>%
  separate(type, sep="_", into=c("layer", "protein")) %>%
  dplyr::select(-protein) %>%
  mutate(layer = as.factor(layer),
        mouse = fct_reorder(as.factor(mouse), perc,
                              .fun=max, .desc=FALSE),
         time = as.factor(time)) %>%
  filter(layer != "total")
head(anovpercs, 2)
## # A tibble: 2 x 14
   mouse light time sex
                             age GCL_chat_mm GCL_bendo_mm INL_chat_mm
```

```
## <fct> <fct> <fct> <chr> <dbl>
                                       <dbl>
                                                   <dbl>
                                                               <dbl>
## 1 11
          light day F
                              95
                                        848.
                                                    26.6
                                                              1071.
                              95
## 2 11
          light day
                    F
                                       848.
                                                    26.6
                                                              1071.
## # ... with 6 more variables: INL_bendo_mm <dbl>, group <fct>,
## # inl_perc_mm <dbl>, gcl_perc_mm <dbl>, layer <fct>, perc <dbl>
Checking Normality
```

#checking if the cell counts are normally distributed shapiro.test(counts\$INL_chat) ## ## Shapiro-Wilk normality test ## data: counts\$INL_chat ## W = 0.89571, p-value = 0.1396 shapiro.test(counts\$INL_bendo) ## ## Shapiro-Wilk normality test ## ## data: counts\$INL_bendo ## W = 0.91934, p-value = 0.2805 shapiro.test(counts\$GCL_chat) ## Shapiro-Wilk normality test ## data: counts\$GCL_chat ## W = 0.88355, p-value = 0.09733 shapiro.test(counts\$GCL_bendo) ## ## Shapiro-Wilk normality test ## ## data: counts\$GCL_bendo ## W = 0.86341, p-value = 0.05396 shapiro.test(counts\$total_chat)

```
##
## Shapiro-Wilk normality test
##
## data: counts$total_chat
## W = 0.88056, p-value = 0.08911
```

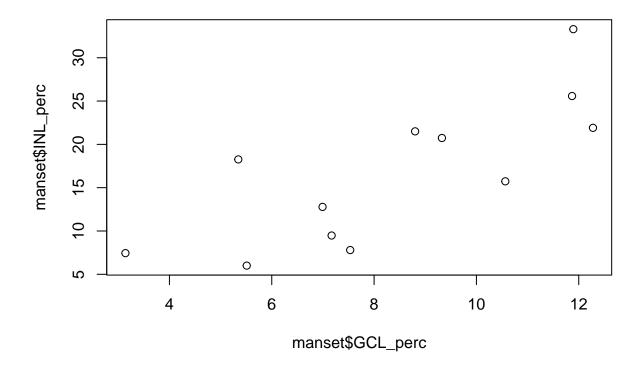
```
shapiro.test(counts$total_bendo)
##
##
   Shapiro-Wilk normality test
## data: counts$total_bendo
## W = 0.90417, p-value = 0.1795
Checking MANOVA Assumptions
manset <- counts %>% dplyr::select(light:time, GCL_perc:INL_perc)
#testing manova assumptions
#multivariate normality
#normal distribution for each combo of indep and dependent variables
anovpercs %>% filter(light == "light" & layer == "GCL") %>% pull(perc) %>% shapiro.test(.)
##
##
   Shapiro-Wilk normality test
##
## data: .
## W = 0.99193, p-value = 0.9934
anovpercs %>% filter(light == "dark" & layer =="GCL") %>% pull(perc) %>% shapiro.test(.)
##
##
   Shapiro-Wilk normality test
##
## data: .
## W = 0.93389, p-value = 0.6104
anovpercs %>% filter(light == "light" & layer =="INL") %>% pull(perc) %>% shapiro.test(.)
##
   Shapiro-Wilk normality test
##
## data:
## W = 0.80407, p-value = 0.06392
anovpercs %>% filter(light == "dark" & layer =="INL") %>% pull(perc) %>% shapiro.test(.)
##
   Shapiro-Wilk normality test
```

data: .

W = 0.91084, p-value = 0.442

```
anovpercs %>% filter(time == "day" & layer =="GCL") %>% pull(perc) %>% shapiro.test(.)
##
##
   Shapiro-Wilk normality test
##
## data:
## W = 0.9171, p-value = 0.4847
anovpercs %>% filter(time == "night" & layer =="GCL") %>% pull(perc) %>% shapiro.test(.)
## Shapiro-Wilk normality test
##
## data: .
## W = 0.96858, p-value = 0.8828
anovpercs %>% filter(time == "day" & layer =="INL") %>% pull(perc) %>% shapiro.test(.)
##
## Shapiro-Wilk normality test
##
## data:
## W = 0.86996, p-value = 0.226
anovpercs %>% filter(time == "night" & layer =="INL") %>% pull(perc) %>% shapiro.test(.)
##
## Shapiro-Wilk normality test
##
## data:
## W = 0.95699, p-value = 0.7962
#all pass
#checking homogenous variance in each combo
#null hyp is that variance is equal
car::leveneTest(manset$GCL_perc, group=manset$light)
## Levene's Test for Homogeneity of Variance (center = median)
        Df F value Pr(>F)
## group 1 0.0151 0.9047
        10
##
car::leveneTest(manset$INL_perc, group=manset$light)
## Levene's Test for Homogeneity of Variance (center = median)
       Df F value Pr(>F)
## group 1 0.0504 0.8268
##
        10
```

```
car::leveneTest(manset$GCL_perc, group=manset$time)
## Levene's Test for Homogeneity of Variance (center = median)
       Df F value Pr(>F)
## group 1 2.5789 0.1394
##
        10
car::leveneTest(manset$INL_perc, group=manset$time)
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 1 1.3768 0.2678
##
        10
#checking for multicollinearity
#low values are good
inllm <- lm(INL_perc ~ light * time, manset)</pre>
car::vif(inllm)
##
       light
                  time light:time
##
gcllm <- lm(GCL_perc ~ light * time, manset)</pre>
car::vif(gcllm)
##
       light
                   time light:time
##
#linearity of dep variables
plot(manset$GCL_perc, manset$INL_perc)
```



#good to go on manova!

MANOVA

```
mantest <- manova(cbind(GCL_perc, INL_perc) ~ light * time, data=manset)</pre>
summary(mantest)
##
              Df Pillai approx F num Df den Df Pr(>F)
                           2.0965
                                       2
                                              7 0.19343
## light
               1 0.37461
## time
               1 0.11699
                           0.4637
                                       2
                                              7 0.64697
                           6.8605
                                       2
                                              7 0.02241 *
## light:time
              1 0.66218
## Residuals
               8
## ---
## Signif. codes:
                   0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
summary.aov(mantest)
##
   Response GCL_perc :
##
               Df Sum Sq Mean Sq F value
                                           Pr(>F)
## light
                1 10.693 10.693 3.0196 0.120464
                1 0.167 0.167 0.0471 0.833664
## time
```

```
## light:time
               1 55.374 55.374 15.6365 0.004211 **
## Residuals
               8 28.331
                          3.541
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
##
   Response INL_perc :
##
              Df Sum Sq Mean Sq F value Pr(>F)
               1 179.82 179.818 4.2818 0.07231 .
## light
## time
               1 21.74 21.741 0.5177 0.49231
## light:time
              1 233.58 233.584 5.5621 0.04607 *
## Residuals
               8 335.96 41.996
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
#light:time interaction is significant overall and for GCL and INL
#look at comparisons through univariate models for each dep variable
summary(gcllm)
##
## Call:
## lm(formula = GCL_perc ~ light * time, data = manset)
## Residuals:
                 1Q
                      Median
                                           Max
## -2.28500 -1.18579 0.03093 0.96322 2.41406
##
## Coefficients:
                      Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                         5.395
                                    1.086
                                            4.965 0.00110 **
## lightdark
                         6.184
                                    1.537
                                            4.025 0.00382 **
## timenight
                         4.061
                                    1.537
                                            2.643 0.02959 *
## lightdark:timenight
                        -8.593
                                    2.173 -3.954 0.00421 **
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 1.882 on 8 degrees of freedom
## Multiple R-squared: 0.7004, Adjusted R-squared: 0.5881
## F-statistic: 6.234 on 3 and 8 DF, p-value: 0.01728
summary(inllm)
##
## Call:
## lm(formula = INL_perc ~ light * time, data = manset)
## Residuals:
##
               1Q Median
                               3Q
      Min
                                      Max
## -9.1204 -2.4874 0.5396 2.6011 9.6505
##
## Coefficients:
##
                      Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                         7.082
                                    3.741
                                            1.893 0.0950 .
                        16.566
                                    5.291
                                            3.131
                                                  0.0140 *
## lightdark
```

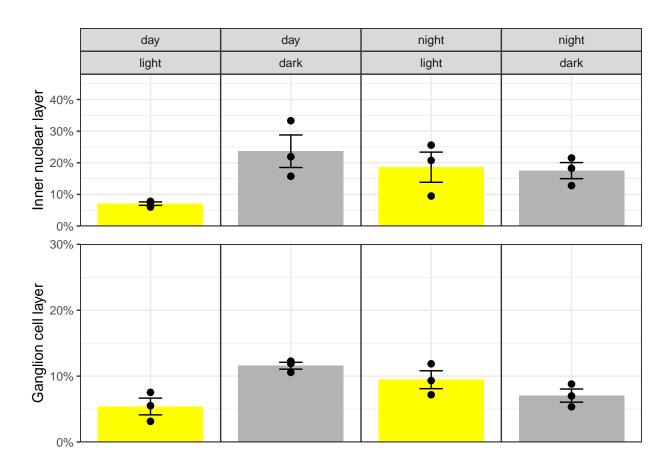
```
5.291 2.176 0.0612 .
## timenight
                       11.516
## lightdark:timenight -17.648
                                  7.483 -2.358 0.0461 *
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 6.48 on 8 degrees of freedom
## Multiple R-squared: 0.5643, Adjusted R-squared: 0.4009
## F-statistic: 3.454 on 3 and 8 DF, p-value: 0.07147
emmeans::emmeans(gcllm, pairwise ~ light * time, adjust="none")$contrasts
## contrast
                            estimate
                                      SE df t.ratio p.value
## light day - dark day
                             -6.18 1.54 8 -4.025 0.0038
## light day - light night
                            -4.06 1.54 8 -2.643 0.0296
## light day - dark night
                            -1.65 1.54 8 -1.075 0.3136
## dark day - light night 2.12 1.54 8 1.382 0.2043 ## dark day - dark night 4.53 1.54 8 2.950 0.0184
## light night - dark night 2.41 1.54 8 1.567 0.1557
emmeans::emmeans(inllm, pairwise ~ light * time, adjust="none")$contrasts
## contrast
                           estimate
                                      SE df t.ratio p.value
## light day - dark day
                           -16.57 5.29 8 -3.131 0.0140
## light day - light night -11.52 5.29 8 -2.176 0.0612
## light day - dark night -10.43 5.29 8 -1.972 0.0841
## dark day - light night 5.05 5.29 8 0.954 0.3678
## dark day - dark night 6.13 5.29 8 1.159 0.2799
## dark day - dark night
                             6.13 5.29 8 1.159 0.2799
## light night - dark night 1.08 5.29 8 0.204 0.8431
emmeans::emmeans(gcllm, pairwise ~ light | time, adjust="none")$contrasts
## time = day:
## contrast
              estimate SE df t.ratio p.value
## light - dark -6.18 1.54 8 -4.025 0.0038
##
## time = night:
## contrast
             estimate SE df t.ratio p.value
## light - dark 2.41 1.54 8 1.567 0.1557
emmeans::emmeans(gcllm, pairwise ~ time | light, adjust="none")$contrasts
## light = light:
## contrast estimate SE df t.ratio p.value
## day - night -4.06 1.54 8 -2.643 0.0296
## light = dark:
## contrast estimate SE df t.ratio p.value
## day - night 4.53 1.54 8 2.950 0.0184
```

```
emmeans::emmeans(inllm, pairwise ~ light | time, adjust="none")$contrasts
## time = day:
## contrast
                          SE df t.ratio p.value
               estimate
## light - dark -16.57 5.29 8 -3.131 0.0140
##
## time = night:
## contrast
               estimate
                          SE df t.ratio p.value
                   1.08 5.29 8 0.204 0.8431
## light - dark
emmeans::emmeans(inllm, pairwise ~ time | light, adjust="none")$contrasts
## light = light:
## contrast
             estimate SE df t.ratio p.value
## day - night -11.52 5.29 8 -2.176 0.0612
##
## light = dark:
## contrast
             estimate SE df t.ratio p.value
## day - night 6.13 5.29 8 1.159 0.2799
```

Plotting

```
plotting_manset <- counts %>%
  dplyr::select(mouse,light:time, GCL_perc:INL_perc)
plotting manset$group <- c("1", "1", "1", "2", "2", "2", "3", "3", "3", "4", "4", "4")
inlpointmanset <- plotting_manset %>%
  mutate(mouse = fct_reorder(as.factor(mouse), INL_perc,
                      .fun=max, .desc=FALSE))
inlpointplot <- inlpointmanset %% ggplot(aes(x=group, y=INL_perc, fill=light)) +
  stat_summary(fun = 'mean', geom="bar") +
  stat_summary(fun.data = mean_se, geom = "errorbar", width=.2) +
  geom_point(size =2) +
  facet_wrap(time ~ factor(light, c("light", "dark")),
             scales="free_x",
             nrow=1.
             as.table=FALSE) +
  theme_bw() %+replace%
  theme(axis.title.x=element_blank(),
        axis.text.x=element_blank(),
        axis.ticks.x = element_blank(),
        legend.position = "none",
        panel.spacing = unit(0,"pt"),
        plot.background = element_rect(fill = "transparent",colour = NA),
        panel.background = element_rect(fill = "transparent",colour = NA)) +
  labs(y="Inner nuclear layer") +
```

```
scale_y_continuous(labels=function(x) paste0(x,"%"),
                     limits=c(0,40),
                     expand = expansion(mult = c(0, 0.2))) +
  scale_fill_manual(values=c("yellow", "gray70"))
gclpointmanset <- plotting_manset %>%
  mutate(mouse = fct reorder(as.factor(mouse), GCL perc,
                      .fun=max, .desc=FALSE))
gclpointplot <- gclpointmanset %>% ggplot(aes(x=group, y=GCL_perc, fill=light)) +
  stat_summary(fun = 'mean', geom="bar") +
  stat_summary(fun.data = mean_se, geom = "errorbar", width=.2) +
  geom_point(size=2) +
  facet_wrap(time ~ factor(light, c("light", "dark")),
             scales="free_x",
             nrow=1,
             as.table=FALSE) +
  theme_bw() %+replace%
  theme(axis.title.x=element_blank(),
        axis.text.x=element_blank(),
        axis.ticks.x = element_blank(),
        legend.position = "none",
        strip.background = element_blank(),
        strip.text.x = element_blank(),
        panel.spacing = unit(0, "pt"),
       plot.background = element_rect(fill = "transparent",colour = NA),
        panel.background = element_rect(fill = "transparent",colour = NA)) +
  labs(y="Ganglion cell layer") +
  scale_y_continuous(labels=function(x) paste0(x,"%"),
                     limits=c(0,30),
                     expand = expansion(mult = c(0, 0))) +
  scale_fill_manual(values=c("yellow", "gray70"))
bothpointplots <- ggarrange(inlpointplot, gclpointplot, nrow=2)</pre>
bothpointplots
```



ggsave("../figures/manova_bendo_ihc.png", plot=bothpointplots, width=6, height=6, bg = "transparent")

Old ANOVA analysis

TukeyHSD(aov)

```
aov <- aov(perc ~ light * time * layer, anovpercs)</pre>
summary(aov)
##
                    Df Sum Sq Mean Sq F value
                                                 Pr(>F)
## light
                        139.1
                                139.1
                                        6.110 0.025056 *
                                  9.0
## time
                          9.0
                                        0.397 0.537298
## layer
                     1
                        417.5
                                417.5 18.337 0.000571 ***
## light:time
                        258.2
                                258.2
                                       11.341 0.003919 **
## light:layer
                         51.4
                                 51.4
                                        2.258 0.152427
                     1
## time:layer
                     1
                         12.9
                                 12.9
                                        0.565 0.463282
                                 30.7
## light:time:layer
                         30.7
                                        1.351 0.262231
                    1
## Residuals
                    16 364.3
                                 22.8
## ---
                   0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
```

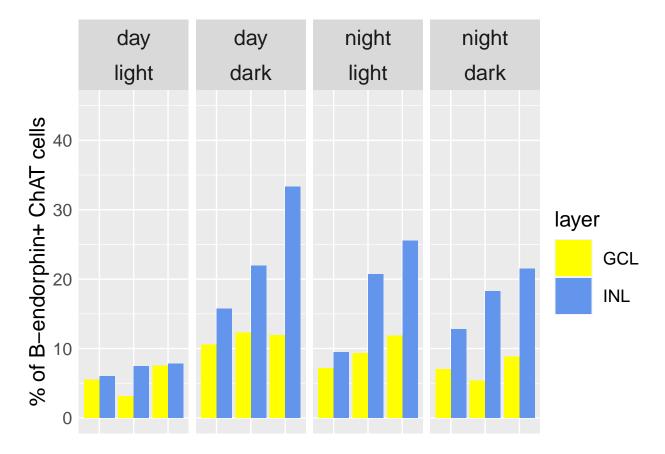
Tukey multiple comparisons of means

```
95% family-wise confidence level
##
##
## Fit: aov(formula = perc ~ light * time * layer, data = anovpercs)
##
## $light
##
                  diff
                             lwr
  dark-light 4.815019 0.6854227 8.944616 0.025056
##
## $time
##
                 diff
                            lwr
  night-day 1.228134 -2.901462 5.357731 0.5372978
##
## $layer
##
               diff
                         lwr
                                  upr
                                           p adj
## INL-GCL 8.341742 4.212145 12.47134 0.0005715
##
## $'light:time'
##
                               diff
                                              lwr
                                                        upr
                                                                p adj
                          11.375105
                                      3.49327511 19.256934 0.0039392
## dark:day-light:day
## light:night-light:day
                           7.788220
                                     -0.09360982 15.670049 0.0533578
## dark:night-light:day
                           6.043154
                                     -1.83867592 13.924983 0.1672187
## light:night-dark:day
                          -3.586885 -11.46871446 4.294945 0.5748113
## dark:night-dark:day
                          -5.331951 -13.21378056 2.549879 0.2528778
## dark:night-light:night -1.745066 -9.62689563 6.136763 0.9197070
##
## $'light:layer'
##
                            diff
                                         lwr
                                                           p adj
                                                   upr
## dark:GCL-light:GCL
                        1.887984 -5.9938453 9.769814 0.9011799
## light:INL-light:GCL
                        5.414707 -2.4671227 13.296536 0.2414734
                       13.156761 5.2749316 21.038591 0.0010656
## dark:INL-light:GCL
## light:INL-dark:GCL
                        3.526723 -4.3551070 11.408552 0.5877993
## dark:INL-dark:GCL
                       11.268777 3.3869473 19.150606 0.0042612
  dark: INL-light: INL
                        7.742054 -0.1397752 15.623884 0.0550899
##
##
  $'time:layer'
##
                             diff
                                        lwr
                                                           p adj
                                                   upr
## night:GCL-day:GCL
                       -0.2357322 -8.117562 7.646097 0.9997654
                        6.8778753 -1.003954 14.759705 0.0986522
## day:INL-day:GCL
## night:INL-day:GCL
                        9.5698762
                                   1.688047 17.451706 0.0149110
## day:INL-night:GCL
                        7.1136075 -0.768222 14.995437 0.0844234
## night:INL-night:GCL
                        9.8056084 1.923779 17.687438 0.0125448
## night:INL-day:INL
                        2.6920009 -5.189829 10.573830 0.7641311
##
##
  $'light:time:layer'
##
                                           diff
                                                        lwr
                                                                          p adj
                                                                  upr
## dark:day:GCL-light:day:GCL
                                    6.18425591
                                                 -7.3043521 19.672864 0.7510493
## light:night:GCL-light:day:GCL
                                    4.06053948
                                                -9.4280686 17.549148 0.9601054
## dark:night:GCL-light:day:GCL
                                     1.65225203 -11.8363560 15.140860 0.9998280
## light:day:INL-light:day:GCL
                                     1.68702656 -11.8015815 15.175635 0.9998025
## dark:day:INL-light:day:GCL
                                    18.25297992
                                                  4.7643719 31.741588 0.0047085
## light:night:INL-light:day:GCL
                                                -0.2856816 26.691535 0.0573262
                                    13.20292650
## dark:night:INL-light:day:GCL
                                    12.12108175 -1.3675263 25.609690 0.0950936
## light:night:GCL-dark:day:GCL
                                    -2.12371644 -15.6123245 11.364892 0.9991136
## dark:night:GCL-dark:day:GCL
                                    -4.53200388 -18.0206119 8.956604 0.9313412
```

```
## dark:day:INL-dark:day:GCL
                                   12.06872400 -1.4198841 25.557332 0.0973977
## light:night:INL-dark:day:GCL
                                   7.01867059 -6.4699375 20.507279 0.6278714
## dark:night:INL-dark:day:GCL
                                   5.93682583 -7.5517822 19.425434 0.7848034
## dark:night:GCL-light:night:GCL -2.40828744 -15.8968955 11.080321 0.9980338
## light:day:INL-light:night:GCL
                                   -2.37351292 -15.8621210 11.115095 0.9982051
## dark:day:INL-light:night:GCL
                                   14.19244044 0.7038324 27.681049 0.0355493
## light:night:INL-light:night:GCL 9.14238703 -4.3462210 22.630995 0.3281466
## dark:night:INL-light:night:GCL
                                    8.06054227 -5.4280658 21.549150 0.4713661
                                   0.03477453 -13.4538335 13.523383 1.0000000
## light:day:INL-dark:night:GCL
## dark:day:INL-dark:night:GCL
                                   16.60072789 3.1121198 30.089336 0.0107646
## light:night:INL-dark:night:GCL 11.55067447 -1.9379336 25.039283 0.1230287
## dark:night:INL-dark:night:GCL
                                   10.46882971 -3.0197783 23.957438 0.1959052
## dark:day:INL-light:day:INL
                                   16.56595336 3.0773453 30.054561 0.0109536
## light:night:INL-light:day:INL
                                  11.51589995 -1.9727081 25.004508 0.1249448
## dark:night:INL-light:day:INL
                                   10.43405519 -3.0545529 23.922663 0.1987393
## light:night:INL-dark:day:INL
                                  -5.05005341 -18.5386615 8.438555 0.8875683
## dark:night:INL-dark:day:INL
                                  -6.13189817 -19.6205062 7.356710 0.7583366
## dark:night:INL-light:night:INL -1.08184476 -14.5704528 12.406763 0.9999901
#set up a table of p-values for labeling the graphs below
label <- tibble(mouse = c("97", "93", "57"),
                perc = c(Inf, Inf, Inf),
                time = c("day", "night", "night"),
                light = c("dark", "light", "dark"),
                label = c("p=0.00009", "p=0.0058", "p=0.037"))
#1 row facet wrap, small
bendo_ihc <- anovpercs %>% ggplot(aes(x=mouse, y=perc)) +
  geom_bar(aes(fill=layer), stat="identity", position="dodge") +
  facet_wrap(time ~ factor(light, c("light", "dark")), scales="free_x", nrow=1, as.table=FALSE) +
  labs(y="% of B-endorphin+ ChAT cells",
       col="Light Condition", shape="Cell Layer",
       title = "Figure 2") +
  theme(axis.title.x=element blank(),
       axis.text.x=element blank(),
       text = element text(size=4),
        plot.title = element_text(size=4, margin=margin(0,0,0,0)), face = "bold"),
        legend.key.size = unit(0.25, 'cm'),
        strip.text.x = element_text(size = 3, margin = margin(0, 0, 0.02, 0, "cm")),
        axis.ticks = element blank(),
        legend.margin=margin(0,0,0,0)) +
  scale_fill_manual(values = c("yellow", "cornflowerblue", "grey50")) +
  geom_text(aes(label=label), data=label, vjust="top", hjust="left", size=1)
#ggsave(filename="../figures/bendo_ihc.png", plot=bendo_ihc, height=1, width=2.5)
#1 row facet wrap, large
bendo_ihc_large <- anovpercs %>% ggplot(aes(x=mouse, y=perc)) +
  geom_bar(aes(fill=layer), stat="identity", position="dodge") +
  facet_wrap(time ~ factor(light, c("light", "dark")), scales="free_x", nrow=1, as.table=FALSE) +
  labs(y="% of B-endorphin+ ChAT cells",
       col="Light Condition", shape="Cell Layer") +
```

-4.49722936 -17.9858374 8.991379 0.9338179

light:day:INL-dark:day:GCL



 $\#ggsave(filename="../figures/bendo_ihc_large_nostats.png", plot=bendo_ihc_large, height=5, width=12)$

```
#2 BY 2 FACET WRAP
bendo_ihc_boxy <- anovpercs %>% ggplot(aes(x=mouse, y=perc)) +
    geom_bar(aes(fill=layer), stat="identity", position="dodge") +
    facet_wrap(time ~ factor(light, c("light", "dark")), scales="free_x", nrow=2, as.table=TRUE) +
    labs(y="% of B-endorphin+ ChAT cells",
        col="Light Condition", shape="Cell Layer") +
    theme(axis.title.x=element_blank(),
        axis.text.x=element_blank(),
        text = element_text(size=4),
        plot.title = element_text(size=4, margin=margin(0,0,0,0), face = "bold"),
```

```
legend.key.size = unit(0.25, 'cm'),
    strip.text.x = element_text(size = 3, margin = margin(0, 0, 0.02, 0, "cm")),
    axis.ticks = element_blank(),
    legend.margin=margin(0,0,0,0)) +
    scale_fill_manual(values = c("yellow", "cornflowerblue", "grey50")) +
    geom_text(aes(label=label), data=label, vjust="top", hjust="left", size=1)

#ggsave(filename="../figures/bendo_ihc_boxy.png", plot=bendo_ihc_boxy, height=4, width=4)
```