Analysis of B-endorphin/ChAT IHC Count Data

C-T Berezin

11/21/2021

```
library(knitr)
library(tidyverse)
library(dplyr)
library(WebPower)
library(ggpubr)
library(car)
counts <- read_csv("../data/B-endo-totals.csv")</pre>
## Rows: 12 Columns: 12
## Delimiter: ","
## chr (3): Light, Time, Sex
## dbl (9): Mouse, ChAT, Bendo, Percentage, Age, GCLchat, GCLbendo, INLchat, IN...
##
## 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.character(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)
#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
#data setup for previous anova analysis
#also used in following 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, 3)
## # A tibble: 3 x 7
    mouse light time sex
                               age layer perc
    <fct> <fct> <fct> <chr> <dbl> <fct> <dbl>
          light day
                               95 GCL
## 1 11
                     F
                                95 INL
                                          7.44
## 2 11
           light day
                     F
## 3 5
           light day
                      Μ
                               141 GCL
                                          7.53
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
```

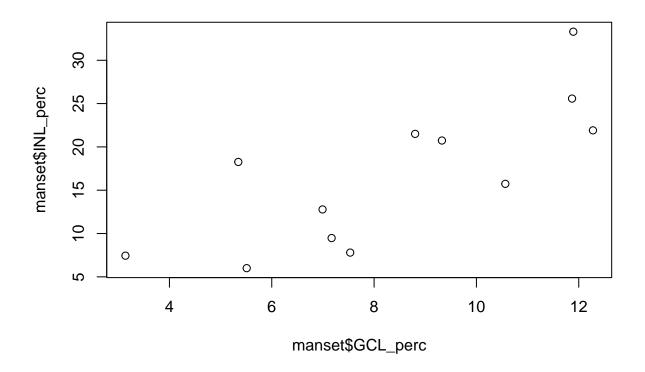
```
anovpercs %>% filter(light == "light" & layer =="INL") %>% pull(perc) %>% shapiro.test(.)
##
##
   Shapiro-Wilk normality test
## data: .
## W = 0.80407, p-value = 0.06392
anovpercs %>% filter(light == "dark" & layer =="INL") %>% pull(perc) %>% shapiro.test(.)
##
  Shapiro-Wilk normality test
##
##
## 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
anovpercs %>% filter(light == "light" & layer == "GCL") %>% pull(perc) %>% shapiro.test(.)
## Shapiro-Wilk normality test
##
## data: .
## W = 0.99193, p-value = 0.9934
#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
           2
                      2
##
```

```
gcllm <- lm(GCL_perc ~ light * time, manset)
car::vif(gcllm)

## light time light:time
## 2 2 3

#linearity of dep variables
plot(manset$GCL_perc, manset$INL_perc)</pre>
```



#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)
## light
               1 0.37461
                           2.0965
                                       2
                                              7 0.19343
               1 0.11699
                           0.4637
                                       2
                                              7 0.64697
## time
## light:time 1 0.66218
                           6.8605
                                       2
                                              7 0.02241 *
## 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
              1 55.374 55.374 15.6365 0.004211 **
## light:time
## 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
               1 21.74 21.741 0.5177 0.49231
## time
## light:time 1 233.58 233.584 5.5621 0.04607 *
             8 335.96 41.996
## Residuals
## Signif. codes: 0 '***' 0.001 '**' 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:
       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 **
                         4.061
                                    1.537
                                            2.643 0.02959 *
## timenight
## lightdark:timenight
                        -8.593
                                    2.173 -3.954 0.00421 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 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:
```

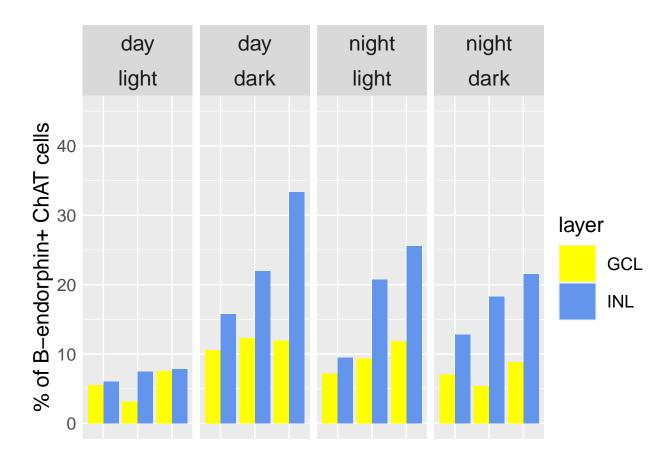
```
1Q Median
##
                                3Q
## -9.1204 -2.4874 0.5396 2.6011 9.6505
##
## Coefficients:
##
                       Estimate Std. Error t value Pr(>|t|)
                          7.082
                                     3.741
                                             1.893 0.0950 .
## (Intercept)
## lightdark
                         16.566
                                     5.291
                                             3.131
                                                     0.0140 *
                                                     0.0612 .
                                             2.176
## timenight
                         11.516
                                     5.291
## 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
plotting manset <- counts %>%
  dplyr::select(mouse,light:time, GCL_perc:INL_perc)
gclplotmanset <- plotting_manset %>%
  mutate(mouse = fct_reorder(as.factor(mouse), GCL_perc,
                      .fun=max, .desc=FALSE))
gclplot <- gclplotmanset %>% ggplot(aes(x=mouse, y=GCL_perc)) +
  geom_bar(stat="identity") +
  facet_wrap(time ~ factor(light, c("light", "dark")), scales="free_x", nrow=1, as.table=FALSE) +
  expand_limits(y=c(0,20)) +
  theme(axis.title.x=element_blank(),
        axis.text.x=element_blank(),
        axis.ticks.x = element_blank()) +
  labs(y="Ganglion cell layer") +
  scale_y_continuous(labels=function(x) paste0(x,"%"))
inlplotmanset <- plotting_manset %>%
  mutate(mouse = fct_reorder(as.factor(mouse), INL_perc,
                      .fun=max, .desc=FALSE))
inlplot <- inlplotmanset %>% ggplot(aes(x=mouse, y=INL_perc)) +
  geom_bar(stat="identity") +
  facet_wrap(time ~ factor(light, c("light", "dark")), scales="free_x", nrow=1, as.table=FALSE) +
  scale_y_continuous(expand = expansion(mult = c(0, 0.3)),
                     labels=function(x) paste0(x,"%")) +
  theme(axis.title.x=element_blank(),
        axis.text.x=element_blank(),
        axis.ticks.x = element blank()) +
  labs(y="Inner nuclear layer")
bothplots <- ggarrange(inlplot, gclplot, nrow=2)</pre>
bothplots <- annotate_figure(bothplots, top="% of ChAT cells expressing \U03B2-endorphin")
ggsave("../figures/manova_bendo_ihc.png", plot=bothplots, width=6, height=6)
```

```
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
                                  9.0
                                        0.397 0.537298
## time
                     1
## layer
                        417.5
                                417.5
                                       18.337 0.000571 ***
                     1
## light:time
                        258.2
                                258.2
                                       11.341 0.003919 **
                     1
## light:layer
                         51.4
                                 51.4
                                        2.258 0.152427
                     1
## time:layer
                         12.9
                                 12.9
                                        0.565 0.463282
                     1
                                 30.7
                                        1.351 0.262231
## light:time:layer 1
                         30.7
## Residuals
                    16 364.3
                                 22.8
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
TukeyHSD(aov)
##
     Tukey multiple comparisons of means
##
       95% family-wise confidence level
##
## Fit: aov(formula = perc ~ light * time * layer, data = anovpercs)
##
## $light
                  diff
                             lwr
                                      upr
                                             p adj
## dark-light 4.815019 0.6854227 8.944616 0.025056
##
## $time
##
                 diff
                            lwr
                                     upr
                                             p adj
## night-day 1.228134 -2.901462 5.357731 0.5372978
##
## $layer
##
               diff
                         lwr
                                  upr
## INL-GCL 8.341742 4.212145 12.47134 0.0005715
##
## $'light:time'
##
                               diff
                                             lwr
                                                        upr
## dark:day-light:day
                          11.375105
                                      3.49327511 19.256934 0.0039392
## 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
                                                           p adj
                                        lwr
                                                  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
## dark:INL-light:GCL 13.156761 5.2749316 21.038591 0.0010656
## 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
                                                          p adj
                                        lwr
                                                  upr
## night:GCL-day:GCL
                       -0.2357322 -8.117562
                                            7.646097 0.9997654
## day:INL-day:GCL
                        6.8778753 -1.003954 14.759705 0.0986522
## night:INL-day:GCL
                        9.5698762 1.688047 17.451706 0.0149110
                        7.1136075 -0.768222 14.995437 0.0844234
## day:INL-night:GCL
## night:INL-night:GCL
                        9.8056084 1.923779 17.687438 0.0125448
                        2.6920009 -5.189829 10.573830 0.7641311
## night:INL-day:INL
##
## $'light:time:layer'
                                          diff
                                                       lwr
                                                                  upr
                                                                          p adj
## 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
                                   13.20292650
                                               -0.2856816 26.691535 0.0573262
## dark:night:INL-light:day:GCL
                                   12.12108175 -1.3675263 25.609690 0.0950936
                                   -2.12371644 -15.6123245 11.364892 0.9991136
## light:night:GCL-dark:day:GCL
## dark:night:GCL-dark:day:GCL
                                   -4.53200388 -18.0206119 8.956604 0.9313412
## light:day:INL-dark:day:GCL
                                   -4.49722936 -17.9858374 8.991379 0.9338179
## 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
                                                 0.7038324 27.681049 0.0355493
                                   14.19244044
## 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
## light:day:INL-dark:night:GCL
                                    0.03477453 -13.4538335 13.523383 1.0000000
## 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
                                                -1.9727081 25.004508 0.1249448
## light:night:INL-light:day:INL
                                   11.51589995
## 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") +
  theme(axis.title.x=element_blank(),
       axis.text.x=element blank(),
       text = element text(size=16),
       legend.key.size = unit(1, 'cm'),
       strip.text.x = element_text(size = 16, margin = margin(0.2, 0, 0.2, 0, "cm")),
       axis.ticks = element_blank(),
        legend.margin=margin(0,0,0,0)) +
  scale_fill_manual(values = c("yellow", "cornflowerblue", "grey50")) +
  expand_limits(y=c(0,45))
  #qeom_text(aes(label=label), data=label, vjust="top", hjust="left", size=6)
bendo_ihc_large
```



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)
```