morphine_regression

nbergum

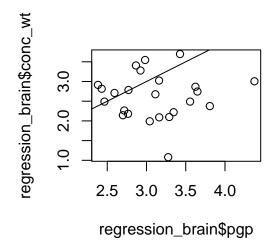
7/25/2022

```
regression <- read.csv("../data/morphine__sex_diff_transporters.csv", fileEncoding = 'UTF-8-BOM')
regression$conc_wt <- (regression$raw_conc/regression$tissue_weight)</pre>
str(regression)
## 'data.frame':
                  48 obs. of 11 variables:
## $ animal
                  : chr "1M" "2M" "4M" "5M" ...
                  : num 56.6 22.1 49.5 87.4 36.6 ...
## $ raw conc
## $ tissue_weight: num
                         20.1 10.3 15.1 31.4 14.7 14.3 20.8 15.7 43.7 14.6 ...
                : num 27.2 30.1 25.8 26.1 27.3 26.2 23.7 27.2 21.3 22.9 ...
## $ body_wt
## $ dose
                  : num 20.6 18.6 21.7 20 19.1 19.9 22 19.2 20.4 19 ...
## $ pgp
                  : num 2.43 2.7 2.92 2.77 2.46 ...
                  : num 1.037 1.249 1.151 0.986 1.205 ...
## $ bcrp
## $ stage
                  : chr "male" "male" "male" ...
                         "male" "male" "male" ...
## $ group
                  : chr
## $ tissue
                  : chr
                         "brain" "brain" "brain" ...
## $ conc_wt
                  : num 2.82 2.14 3.28 2.78 2.49 ...
regression <- regression %>% mutate(stage = as.factor(stage),
                                   group = as.factor(group),
                                   tissue = factor(as.factor(tissue), c("retina", "brain"))
head(regression)
    animal raw_conc tissue_weight body_wt dose
##
                                                   pgp
                                                            bcrp stage group
## 1
        1M 56.5895
                             20.1
                                     27.2 20.6 2.429027 1.0366970
                                                                  male
                                                                       male
## 2
        2M 22.0798
                             10.3
                                     30.1 18.6 2.699340 1.2487060
                                                                  male
                                                                        male
## 3
        4M 49.5114
                             15.1
                                     25.8 21.7 2.923597 1.1505545
                                                                  male
                                                                        male
## 4
                                    26.1 20.0 2.770560 0.9860451
        5M 87.4161
                             31.4
                                                                  male
                                                                        male
                                     27.3 19.1 2.464103 1.2046594
                             14.7
## 5
        6M 36.5960
                                                                  male male
                             14.3
        7M 41.6730
                                     26.2 19.9 2.381759 1.1407836 male male
## 6
##
    tissue conc wt
## 1 brain 2.815398
## 2 brain 2.143670
## 3 brain 3.278901
## 4 brain 2.783952
## 5 brain 2.489524
## 6 brain 2.914196
```

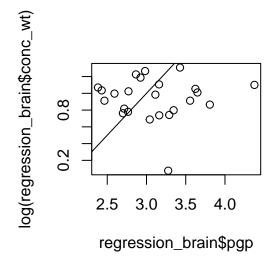
```
pgp_plot <- ggplot(regression, aes(pgp, conc_wt, col=tissue)) +</pre>
  geom_point() +
  geom_smooth(method = "lm", se = T) +
  xlim(0,5) +
 theme_bw()
bcrp_plot <- ggplot(regression, aes(bcrp, conc_wt, col=tissue)) +</pre>
  geom point() +
  geom_smooth(method = "lm", se = T) +
  xlim(0,5) +
 ylab("") +
 theme_bw() %+replace%
  theme(axis.text.y=element_blank())
combined_correlation_plots <- ggarrange(pgp_plot, bcrp_plot, ncol=2, common.legend = TRUE, legend="bott
## 'geom_smooth()' using formula 'y ~ x'
## 'geom_smooth()' using formula 'y ~ x'
## 'geom_smooth()' using formula 'y ~ x'
ggsave("../figures/combined_correlation_plots.png", plot=combined_correlation_plots, width=6, height =4
regression_brain <- subset(regression, tissue == "brain")</pre>
head(regression_brain)
     animal raw_conc tissue_weight body_wt dose
##
                                                     pgp
                                                              bcrp stage group
## 1
        1M 56.5895
                              20.1
                                     27.2 20.6 2.429027 1.0366970 male male
## 2
        2M 22.0798
                              10.3
                                     30.1 18.6 2.699340 1.2487060 male male
        4M 49.5114
                                     25.8 21.7 2.923597 1.1505545 male
## 3
                              15.1
                                                                         male
## 4
        5M 87.4161
                             31.4 26.1 20.0 2.770560 0.9860451 male male
## 5
        6M 36.5960
                             14.7 27.3 19.1 2.464103 1.2046594 male male
## 6
        7M 41.6730
                             14.3
                                     26.2 19.9 2.381759 1.1407836 male male
   tissue conc_wt
##
## 1 brain 2.815398
## 2 brain 2.143670
## 3 brain 3.278901
## 4 brain 2.783952
## 5 brain 2.489524
## 6 brain 2.914196
shapiro.test(regression_brain$conc_wt)
##
## Shapiro-Wilk normality test
## data: regression_brain$conc_wt
## W = 0.9661, p-value = 0.5723
shapiro.test(regression_brain$pgp)
```

```
##
## Shapiro-Wilk normality test
##
## data: regression_brain$pgp
## W = 0.96357, p-value = 0.5143

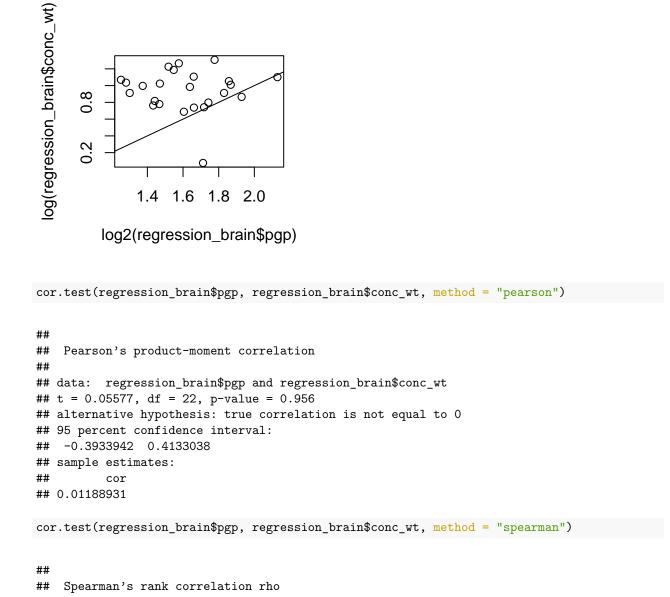
plot(regression_brain$pgp, regression_brain$conc_wt)
abline(0,1)
```



plot(regression_brain\$pgp, log(regression_brain\$conc_wt))
abline(-2,1)



plot(log2(regression_brain\$pgp), log(regression_brain\$conc_wt))
abline(-1,1)



```
shapiro.test(regression_brain$bcrp)
```

S = 2306, p-value = 0.9919

rho

sample estimates:

-0.002608696

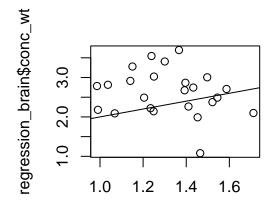
##

data: regression_brain\$pgp and regression_brain\$conc_wt

alternative hypothesis: true rho is not equal to 0

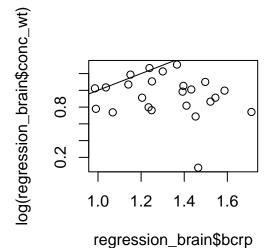
```
##
## Shapiro-Wilk normality test
##
## data: regression_brain$bcrp
## W = 0.9744, p-value = 0.7749
```

```
plot(regression_brain$bcrp, regression_brain$conc_wt)
abline(1,1)
```



regression_brain\$bcrp

plot(regression_brain\$bcrp, log(regression_brain\$conc_wt))
abline(0,1)

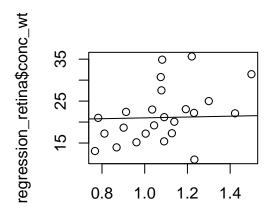


plot(log2(regression_brain\$bcrp), log(regression_brain\$conc_wt))
abline(0.5,1)

```
log(regression_brain$conc_wt)
                         000
                     00
                                 O
       0.2
                    0.2
                           0.4
                                  0.6
                                        8.0
          log2(regression_brain$bcrp)
cor.test(regression_brain$bcrp, regression_brain$conc_wt, method = "pearson")
##
```

```
##
   Pearson's product-moment correlation
##
## data: regression_brain$bcrp and regression_brain$conc_wt
## t = -0.86512, df = 22, p-value = 0.3963
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
  -0.5449099 0.2395396
## sample estimates:
##
          cor
## -0.1813841
cor.test(regression_brain$bcrp, regression_brain$conc_wt, method = "spearman")
##
##
   Spearman's rank correlation rho
## data: regression_brain$bcrp and regression_brain$conc_wt
## S = 2736, p-value = 0.3733
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
##
          rho
## -0.1895652
regression_retina <- subset(regression, tissue == "retina")</pre>
head(regression_retina)
##
      animal raw_conc tissue_weight body_wt dose
                                                         pgp
                                                                  bcrp stage group
                                       27.2 20.55 1.0782765 1.2298197
## 25
          1M 104.7333
                                3.8
```

```
99.2028
                                4.3
                                       26.1 20.00 1.1925447 0.9961628
## 28
          5M
                                7.0
                                       27.3 19.12 0.8683951 0.8810145
## 29
          6M 97.7141
                                                                       male
                                                                              male
                                5.2
                                       26.2 19.92 1.0440075 1.1553400 male
## 30
          7M 99.8877
##
      tissue conc_wt
## 25 retina 27.56139
## 26 retina 17.24694
## 27 retina 22.08005
## 28 retina 23.07042
## 29 retina 13.95916
## 30 retina 19.20917
shapiro.test(regression_retina$conc_wt)
##
##
   Shapiro-Wilk normality test
## data: regression_retina$conc_wt
## W = 0.94676, p-value = 0.2303
shapiro.test(regression_retina$pgp)
##
   Shapiro-Wilk normality test
##
##
## data: regression_retina$pgp
## W = 0.973, p-value = 0.7409
```

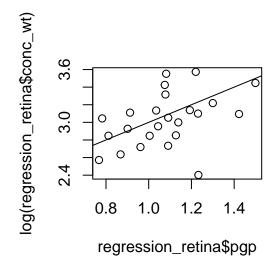


regression_retina\$pgp

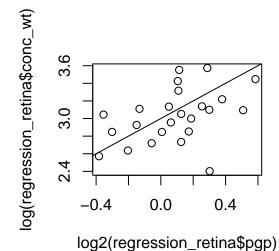
abline(20,1)

plot(regression_retina\$pgp, regression_retina\$conc_wt)

```
plot(regression_retina$pgp, log(regression_retina$conc_wt))
abline(2,1)
```



```
plot(log2(regression_retina$pgp), log(regression_retina$conc_wt))
abline(3,1)
```

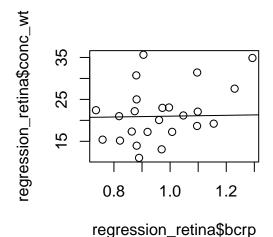


cor.test(regression_retina\$pgp, regression_retina\$conc_wt, method = "pearson")

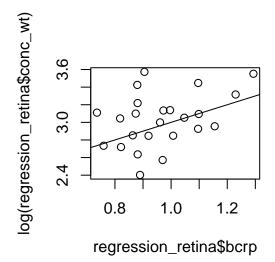
```
##
## Pearson's product-moment correlation
##
## data: regression_retina$pgp and regression_retina$conc_wt
## t = 2.2598, df = 22, p-value = 0.03408
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.03714565 0.71265643
## sample estimates:
## cor
```

abline(20,1)

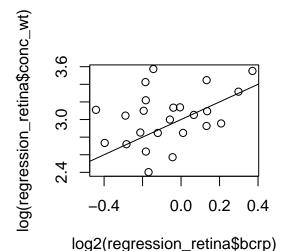
```
cor.test(regression_retina$pgp, regression_retina$conc_wt, method = "spearman")
##
##
    Spearman's rank correlation rho
##
## data: regression_retina$pgp and regression_retina$conc_wt
## S = 1304, p-value = 0.03568
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
##
         rho
## 0.4330435
shapiro.test(regression_retina$bcrp)
##
##
    Shapiro-Wilk normality test
##
## data: regression_retina$bcrp
## W = 0.9579, p-value = 0.3978
plot(regression_retina$bcrp, regression_retina$conc_wt)
```



plot(regression_retina\$bcrp, log(regression_retina\$conc_wt))
abline(2,1)



```
plot(log2(regression_retina$bcrp), log(regression_retina$conc_wt))
abline(3,1)
```



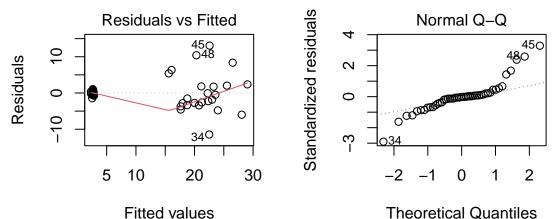
cor.test(regression_retina\$bcrp, regression_retina\$conc_wt, method = "pearson")

```
##
## Pearson's product-moment correlation
##
## data: regression_retina$bcrp and regression_retina$conc_wt
## t = 2.0013, df = 22, p-value = 0.05785
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## -0.01300906 0.68707216
## sample estimates:
## cor
```

```
cor.test(regression_retina$bcrp, regression_retina$conc_wt, method = "spearman")
```

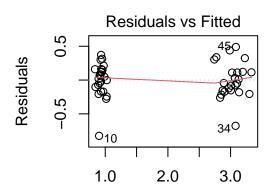
```
##
## Spearman's rank correlation rho
##
## data: regression_retina$bcrp and regression_retina$conc_wt
## S = 1632, p-value = 0.1682
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
## rho
## 0.2904348

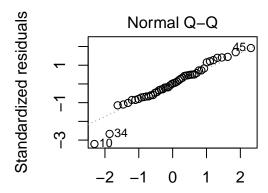
lm_ex1 <- lm(conc_wt ~ pgp*tissue + bcrp*tissue, data = regression)
plot(lm_ex1, which=c(1,2))</pre>
```



Im(conc_wt ~ pgp * tissue + bcrp * tiss Im(conc_wt ~ pgp * tissue + bcrp * tiss

lm_ex <- lm(log(conc_wt) ~ log2(pgp)*tissue + log2(bcrp)*tissue, data = regression)
plot(lm_ex, which=c(1,2))</pre>

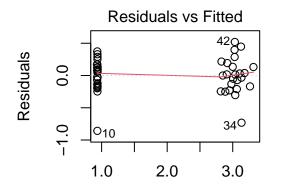




Fitted values Theoretical Quantiles y(conc_wt) ~ log2(pgp) * tissue + log2(bcry(conc_wt) ~ log2(bcry(conc_wt) ~

summary(lm_ex)\$coefficient

```
##
                            Estimate Std. Error
                                                                Pr(>|t|)
                                                    t value
## (Intercept)
                           3.0193972 0.06176213 48.8875210 1.186804e-38
## log2(pgp)
                           0.4385311 0.22416220 1.9563116 5.709991e-02
## tissuebrain
                          -2.1538742 0.43938139 -4.9020608 1.462185e-05
## log2(bcrp)
                           0.4272982 0.27311844 1.5645161 1.252008e-01
## log2(pgp):tissuebrain -0.3335960 0.37326691 -0.8937197 3.765647e-01
## tissuebrain:log2(bcrp) -0.6872715 0.40615462 -1.6921424 9.802618e-02
pgp_subset <- regression %>% select(-bcrp)
bcrp_subset <- regression %>% select(-pgp)
#untransformed
#lm_pgp <- lm(conc_wt ~ pgp * tissue, data=pgp_subset)</pre>
#plot(lm_pgp, which=c(1,2))
#concentration log transformed
lm_pgp_log <- lm(log(conc_wt) ~ pgp * tissue, data=pgp_subset)</pre>
plot(lm_pgp_log, which=c(1,2))
```

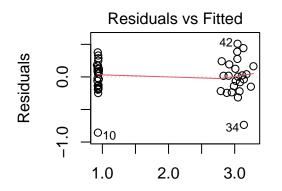


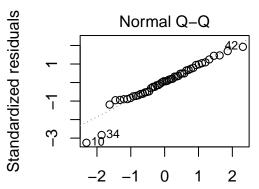
Fitted values Im(log(conc_wt) ~ pgp * tissue)

Theoretical Quantiles Im(log(conc_wt) ~ pgp * tissue)

```
#pgp log transformed
#lm_pgp_log1 <- lm(conc_wt ~ log2(pgp) * tissue, data=pgp_subset)
#plot(lm_pgp_loglog, which=c(1,2))

#both transformed
lm_pgp_loglog <- lm(log(conc_wt) ~ log2(pgp) * tissue, data=pgp_subset)
plot(lm_pgp_loglog, which=c(1,2))</pre>
```





Fitted values Im(log(conc_wt) ~ log2(pgp) * tissue

Theoretical Quantiles Im(log(conc_wt) ~ log2(pgp) * tissue

```
summary(lm_pgp_log)
```

```
##
## Call:
## lm(formula = log(conc_wt) ~ pgp * tissue, data = pgp_subset)
##
## Residuals:
```

```
1Q Median
                                  3Q
## -0.85801 -0.16073 0.01583 0.14377 0.52006
## Coefficients:
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                   2.2967
                              0.3281 7.000 1.14e-08 ***
                   0.6796
                              0.3000 2.265 0.02850 *
## pgp
## tissuebrain
                              0.4906 -2.742 0.00879 **
                  -1.3453
## pgp:tissuebrain -0.6846
                              0.3218 -2.128 0.03901 *
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.2705 on 44 degrees of freedom
## Multiple R-squared: 0.9427, Adjusted R-squared: 0.9388
## F-statistic: 241.4 on 3 and 44 DF, p-value: < 2.2e-16
anova(lm_pgp_log)
## Analysis of Variance Table
## Response: log(conc_wt)
##
            Df Sum Sq Mean Sq F value
                                         Pr(>F)
## pgp
              1 45.719 45.719 624.9245 < 2.2e-16 ***
              1 6.921 6.921 94.6050 1.56e-12 ***
## tissue
## pgp:tissue 1 0.331 0.331
                               4.5267 0.03901 *
## Residuals 44 3.219
                        0.073
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
emmeans::emmeans(lm_pgp_log, pairwise ~ tissue)$contrasts
## NOTE: Results may be misleading due to involvement in interactions
                             SE df t.ratio p.value
## contrast
                  estimate
## retina - brain 2.78 0.335 44 8.294 <.0001
## Results are given on the log (not the response) scale.
```