qPCR Transporter Figures

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```
library(tidyverse)
library(ggplot2)
library(ggthemes)
library(forcats)
library(ggpubr)
#library(writexl)
library(emmeans)
library(lme4)
library(car)
library(lmerTest)
library(svglite)
```

Comparing chronic saline/morphine for Bcrp (rerun of samples for P-gp from Bergum 2022)

Reading in & tidying the data

```
Sample Pgp.RGE Bcrp.RGE Tissue Treatment
##
        C1 1.7701376 1.6160126
## 1
                                 hyp
                                             C
        C1 0.4207643 0.8918919
                                 ret
## 3
        C2 1.7514899 1.6263272
                                 hyp
                                             C
        C2 0.9387229 1.2154637
                                             C
## 4
                                 ret
                                             С
       C3 1.2344328 2.3715652
## 5
                                 hyp
       C3 0.7343906 1.0166798
## 6
                                 ret
```

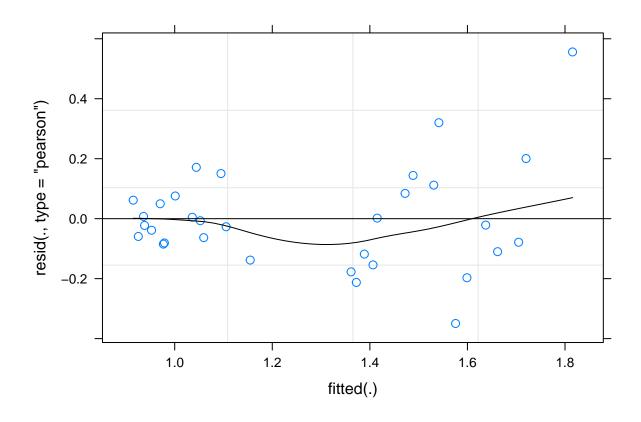
```
transporters_long <- transporters %>% pivot_longer(cols=c("Pgp.RGE", "Bcrp.RGE"), names_to = "gene", va
transporters_long <- transporters_long %>% mutate(gene = str_extract(gene, "[:alpha:]+(?=.)"))
transporters_long <- transporters_long %% mutate(gene = factor(as.factor(gene), c("Pgp", "Bcrp")))
head(transporters_long)
## # A tibble: 6 x 5
##
    Sample Tissue Treatment gene
     <chr>
           <fct> <fct>
                             <fct> <dbl>
## 1 C1
                   С
            hyp
                             Pgp
                                   1.77
## 2 C1
            hyp
                   С
                             Bcrp 1.62
## 3 C1
                   С
                                   0.421
            ret
                             Pgp
## 4 C1
                   С
                             Bcrp 0.892
            ret
## 5 C2
                   С
            hyp
                             Pgp
                                   1.75
## 6 C2
                   С
                                  1.63
            hyp
                             Bcrp
```

Stats on Bcrp only (chronic vs saline)

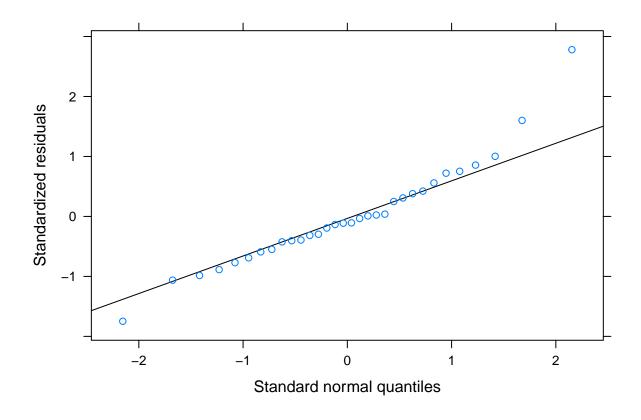
```
#bcrp only
chronic <- transporters %>% filter(Treatment %in% c("C", "S"))
chronic
```

```
##
      Sample
               Pgp.RGE Bcrp.RGE Tissue Treatment
## 1
          C1 1.7701376 1.6160126
                                                  C
                                     hyp
## 2
                                                  С
          C1 0.4207643 0.8918919
                                     ret
## 3
          C2 1.7514899 1.6263272
                                                  С
                                     hyp
## 4
          C2 0.9387229 1.2154637
                                                  C
                                     ret
                                                  С
## 5
          C3 1.2344328 2.3715652
                                     hyp
## 6
          C3 0.7343906 1.0166798
                                                  С
                                     ret
                                                  С
## 7
          C4 1.8650058 1.9206809
                                     hyp
## 8
          C4 0.6869529 0.9963086
                                     ret
                                                  С
## 9
                                                  С
          C5 1.3727992 1.2261280
                                     hyp
          C5 1.2258085 0.9766218
                                                  C
## 10
                                     ret
          C6 1.4958638 1.5518807
                                                  С
## 11
                                     hyp
                                                  С
## 12
          C6 0.7876697 1.0768726
                                     ret
                                                  С
## 13
          C7 2.0301205 1.4020048
                                     hyp
## 14
          C7 1.1768876 0.9157910
                                                  С
                                     ret
                                                  S
## 15
          S1 1.8016888 1.5565636
                                     hyp
                                                  S
## 16
          S1 1.0200728 1.0407493
                                     ret
                                                  S
## 17
          S2 1.3689686 1.8615469
                                     hyp
          S2 0.8222355 1.0784274
                                                  S
## 18
                                     ret
## 19
          S3 1.5582238 1.1842034
                                                  S
                                     hyp
                                                  S
## 20
          S3 1.0081232 0.8660578
                                     ret
## 21
          S4 1.0379047 1.2523247
                                                  S
                                     hyp
## 22
                                                  S
          S4 0.7776565 1.0200962
                                     ret
## 23
          S5 1.6163239 1.2704658
                                                  S
                                     hyp
                                                  S
## 24
          S5 0.8109336 0.9139269
                                     ret
## 25
          S6 1.7206882 1.1597168
                                                  S
                                     hyp
          S6 1.0835666 0.9436934
                                                  S
## 26
                                     ret
```

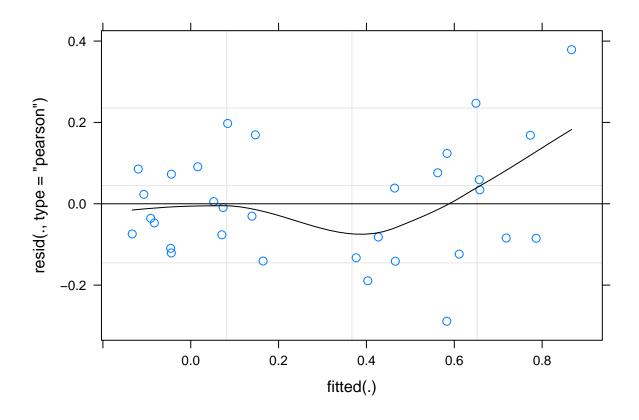
```
## 27
          S7 2.3260412 1.6427723
                                     hyp
                                                  S
## 28
          S7 1.4357874 1.2453351
                                     ret
                                                  S
                                                  S
## 29
          S8 2.0696005 1.6327948
                                     hyp
## 30
          S8 1.0679296 1.0456424
                                     ret
                                                  S
## 31
                                                  S
          S9 2.4339414 1.4168757
                                     hyp
## 32
          S9 1.1287493 0.8979728
                                                  S
                                     ret
c_bcrp_lm <- lmer(Bcrp.RGE ~ Tissue * Treatment + (1|Sample), data=chronic)</pre>
plot(c_bcrp_lm, type=c("p","smooth"), col.line=1)
```



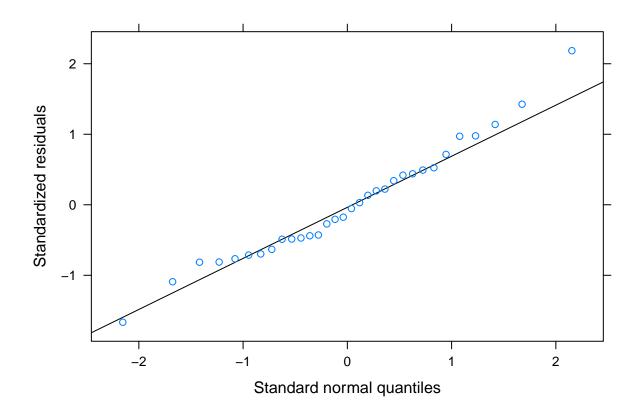
lattice::qqmath(c_bcrp_lm)



```
c_bcrp_log2_lm <- lmer(log2(Bcrp.RGE) ~ Tissue * Treatment + (1|Sample), data=chronic)
plot(c_bcrp_log2_lm, type=c("p","smooth"), col.line=1)</pre>
```



lattice::qqmath(c_bcrp_log2_lm)



```
shapiro.test(chronic$Bcrp.RGE)
##
    Shapiro-Wilk normality test
##
##
## data: chronic$Bcrp.RGE
## W = 0.89092, p-value = 0.003667
shapiro.test(log2(chronic$Bcrp.RGE))
##
    Shapiro-Wilk normality test
##
##
## data: log2(chronic$Bcrp.RGE)
## W = 0.94026, p-value = 0.07615
anova(c_bcrp_log2_lm)
## Type III Analysis of Variance Table with Satterthwaite's method
##
                     Sum Sq Mean Sq NumDF DenDF F value
## Tissue
                    2.89290 2.89290
                                             14 96.2633 1.182e-07 ***
                                        1
## Treatment
                    0.04000 0.04000
                                        1
                                             14
                                                1.3311
                                                            0.2679
## Tissue:Treatment 0.07294 0.07294
                                             14 2.4271
                                        1
                                                            0.1416
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
```

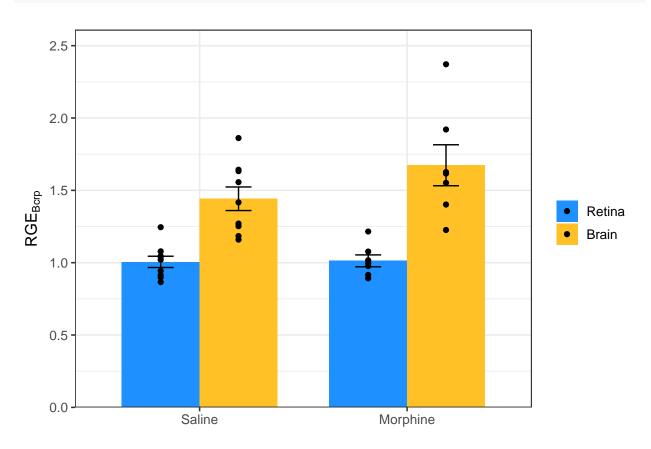
```
emmeans::emmeans(c_bcrp_log2_lm, pairwise ~ Treatment | Tissue)$contrasts
## Tissue = ret:
## contrast estimate
                        SE
                            df t.ratio p.value
          -0.0115 0.112 24.3 -0.103 0.9188
##
## Tissue = hyp:
## contrast estimate
                             df t.ratio p.value
                        SE
## S - C
           -0.2040 0.112 24.3 -1.822 0.0808
##
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
emmeans::emmeans(c_bcrp_log2_lm, pairwise ~ Tissue | Treatment)$contrasts
## Treatment = S:
## contrast estimate
                          SE df t.ratio p.value
   ret - hyp -0.510 0.0817 14 -6.239 <.0001
##
## Treatment = C:
## contrast estimate
                          SE df t.ratio p.value
## ret - hyp -0.702 0.0927 14 -7.580 <.0001
##
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
c_bcrp_sumstats <- chronic %>% group_by(Treatment, Tissue) %>% summarise(
 n = n(),
 mean = mean(Bcrp.RGE),
 sd = sd(Bcrp.RGE),
 se = sd/sqrt(n),
 log2_mean = mean(log2(Bcrp.RGE)),
 log2_sd = sd(log2(Bcrp.RGE)),
 log2_se = log2_sd/sqrt(n)
)
## 'summarise()' has grouped output by 'Treatment'. You can override using the
## '.groups' argument.
c_bcrp_sumstats
## # A tibble: 4 x 9
## # Groups:
              Treatment [2]
    Treatment Tissue
                                           se log2_mean log2_sd log2_se
                        n mean
                                    sd
              <fct> <int> <dbl> <dbl> <dbl>
##
    <fct>
                                                  <dbl>
                                                         <dbl>
                                                                 <dbl>
## 1 S
                        9 1.01 0.117 0.0389 -4.39e-11
                                                         0.163 0.0542
              ret
## 2 S
                         9 1.44 0.244 0.0815 5.10e- 1
                                                         0.242 0.0806
              hyp
                         7 1.01 0.109 0.0411 1.15e- 2 0.150 0.0567
## 3 C
              ret
## 4 C
              hyp
                         7 1.67 0.375 0.142
                                              7.14e- 1
                                                        0.308 0.116
```

Visualizing chronic vs saline Bcrp

```
#visualizing tissue effect per treatment
c_bcrp_tiss_plot <- ggplot() +</pre>
  stat_summary(data=chronic, aes(x=Treatment, y=Bcrp.RGE, fill=Tissue), fun='mean',
               geom="bar", width=0.75, position=position_dodge(width=0.75)) +
  geom_errorbar(data=c_bcrp_sumstats, aes(x=Treatment, ymin=mean-se,
                                           ymax=mean+se, fill=Tissue),
                width=0.25, position=position_dodge(width=0.75)) +
  geom_point(data=chronic, aes(x=Treatment, y=Bcrp.RGE, fill=Tissue),
             position=position_dodge(width=0.75)) +
  #facet_grid(~Tissue, scales="free_y") +
  scale_y_continuous(expand = expansion(mult = c(0, 0.1))) +
  theme bw(base size = 12) %+replace%
   theme(legend.title = element_blank()) +
  scale_fill_manual(values=c("dodgerblue", "goldenrod1"),
                    labels=c("Retina", "Brain")) +
  scale x discrete(labels=c("S" = "Saline", "C" = "Morphine")) +
  xlab("") +
  ylab(expression(RGE[Bcrp]))
```

Warning: Ignoring unknown aesthetics: fill

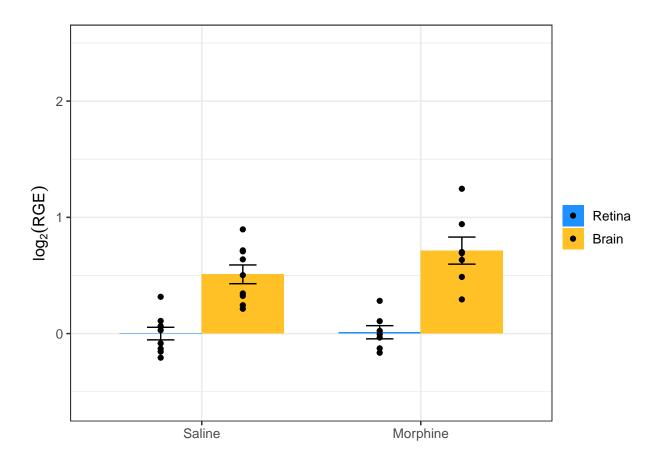
c_bcrp_tiss_plot



```
\#ggsave(filename="../../figures/qRT-PCR/c\_bcrp\_tiss\_plot.png", plot=c\_bcrp\_tiss\_plot, height=4, width=4, widt
\#ggsave(filename=".../../figures/qRT-PCR/c\_bcrp\_tiss\_plot.svg", plot=c\_bcrp\_tiss\_plot, height=4, width=4, wid
c_bcrp_tiss_log_plot <- ggplot() +</pre>
       stat_summary(data=chronic, aes(x=Treatment, y=log2(Bcrp.RGE), fill=Tissue),
                                                              fun = 'mean', geom="bar", width=0.75, position="dodge") +
       geom_errorbar(data=c_bcrp_sumstats, aes(x=Treatment, ymin=log2_mean-log2_se,
                                                                                                                                                                           ymax=log2_mean+log2_se, fill=Tissue),
                                                                  width=0.25, position=position_dodge(width=0.75)) +
       geom_point(data=chronic, aes(x=Treatment, y=log2(Bcrp.RGE), fill=Tissue),
                                                     stat="identity", position=position_dodge(width=0.75)) +
       #facet_grid(~Tissue, scales="free_y") +
       \#scale\_y\_continuous(expand = expansion(mult = c(0.1, 0.2))) +
       ylim(-0.6, 2.5) +
        #scale_fill_manual(values=c("maroon1", "springgreen3")) +
       theme_bw(base_size = 12) %+replace%
                theme(legend.title = element_blank(),
                                         legend.box.spacing = unit(1, "pt"),
                                         axis.title.x=element_blank(),
                                         plot.title = element_text(hjust = 0.5, vjust=1.5)
                                         ) +
       scale_fill_manual(values=c("dodgerblue", "goldenrod1"),
                                                                                   labels=c("Retina", "Brain")) +
       scale_x_discrete(labels=c("S" = "Saline", "C" = "Morphine")) +
       xlab("") +
       ylab(expression(log[2](RGE)))
```

Warning: Ignoring unknown aesthetics: fill

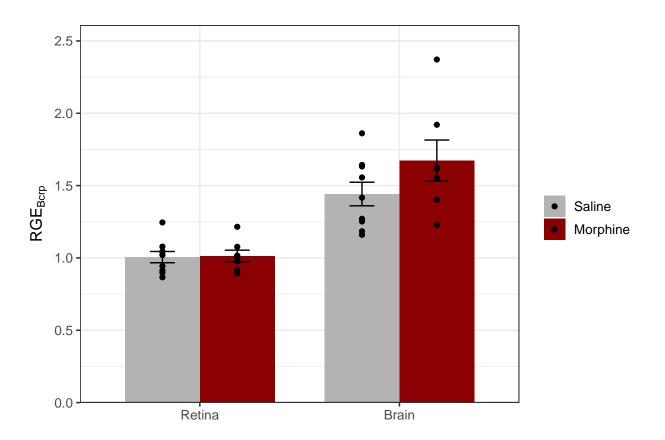
```
c_bcrp_tiss_log_plot
```



```
\#ggsave(filename="../../figures/qRT-PCR/c\_bcrp\_tiss\_log\_plot.svg",\ plot=c\_bcrp\_tiss\_log\_plot,\ height=4,
#visualizing treatment effect per tissue
c_bcrp_trt_plot <- ggplot() +</pre>
 stat_summary(data=chronic, aes(x=Tissue, y=Bcrp.RGE, fill=Treatment),
             fun = 'mean', geom="bar", width=0.75, position="dodge") +
 geom_errorbar(data=c_bcrp_sumstats, aes(x=Tissue, ymin=mean-se,
                                       ymax=mean+se, fill=Treatment),
              width=0.25, position=position_dodge(width=0.75)) +
 geom_point(data=chronic, aes(x=Tissue, y=Bcrp.RGE, fill=Treatment),
           stat="identity", position=position_dodge(width=0.75)) +
 scale_y_continuous(expand = expansion(mult = c(0, 0.1))) +
 theme_bw(base_size = 12) %+replace%
   theme(legend.title = element_blank()) +
 scale_fill_manual(values=c("gray70", "red4"),
                  labels=c("Saline", "Morphine")) +
 scale_x_discrete(labels=c("ret" = "Retina", "hyp" = "Brain")) +
 xlab("") +
 ylab(expression(RGE[Bcrp]))
```

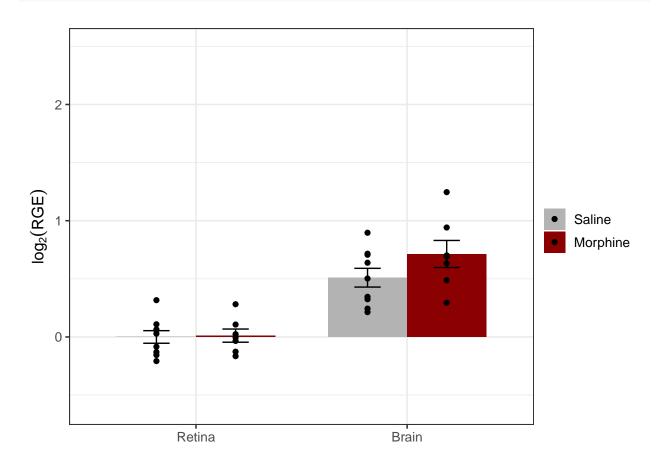
Warning: Ignoring unknown aesthetics: fill

c_bcrp_trt_plot



```
\#ggsave(filename="../../figures/qRT-PCR/c\_bcrp\_trt\_plot.png", plot=c\_bcrp\_trt\_plot, height=4, width=4)
#ggsave(filename="../../figures/qRT-PCR/c_bcrp_trt_plot.svg", plot=c_bcrp_trt_plot, height=4, width=4)
c_bcrp_trt_log_plot <- ggplot() +</pre>
  stat_summary(data=chronic, aes(x=Tissue, y=log2(Bcrp.RGE), fill=Treatment),
               fun = 'mean', geom="bar", width=0.75, position="dodge") +
  geom_errorbar(data=c_bcrp_sumstats, aes(x=Tissue, ymin=log2_mean-log2_se,
                                          ymax=log2_mean+log2_se, fill=Treatment),
                width=0.25, position=position_dodge(width=0.75)) +
  geom_point(data=chronic, aes(x=Tissue, y=log2(Bcrp.RGE), fill=Treatment),
             stat="identity", position=position_dodge(width=0.75)) +
  \#scale\_y\_continuous(expand = expansion(mult = c(0.1, 0.25))) +
  ylim(-0.6, 2.5) +
  theme_bw(base_size = 12) %+replace%
    theme(legend.title = element_blank(),
          legend.box.spacing = unit(1, "pt"),
          axis.title.x=element_blank(),
          plot.title = element_text(hjust = 0.5, vjust=1.5)) +
  scale_fill_manual(values=c("gray70", "red4"),
                    labels=c("Saline", "Morphine")) +
  scale_x_discrete(labels=c("ret" = "Retina", "hyp" = "Brain")) +
  xlab("") +
  ylab(expression(log[2](RGE)))
```

c_bcrp_trt_log_plot



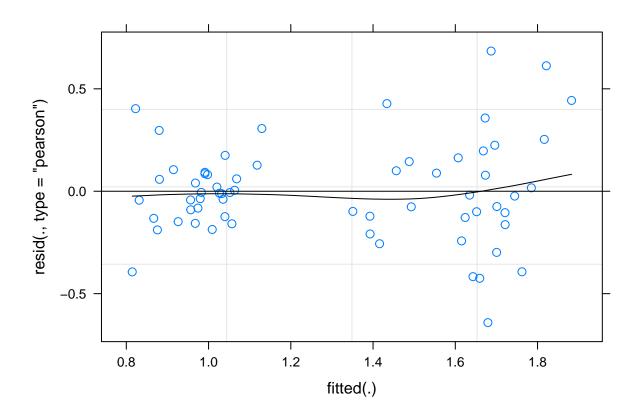
 $\#ggsave(filename="../../figures/qRT-PCR/c_bcrp_trt_log_plot.png", plot=c_bcrp_trt_log_plot, height=4, w\\ \#ggsave(filename="../../figures/qRT-PCR/c_bcrp_trt_log_plot.svg", plot=c_bcrp_trt_log_plot, height=4, w\\ \#ggsave(filename="../../figures/qRT-PCR/c_bcrp_trt_log_plot.svg", plot=c_bcrp_trt_log_plot, height=4, w\\ \#ggsave(filename=".../../figures/qRT-PCR/c_bcrp_trt_log_plot.svg", plot=c_bcrp_trt_log_plot, height=4, w\\ \#ggsave(filename=".../figures/qRT-PCR/c_bcrp_trt_log_plot.svg", plot=c_bcrp_trt_log_plot, height=4, w\\ \#ggsave(filename=".../figures/qRT-PCR/c_bcrp_trt_log_plot, height=4, w\\ \#ggsave(filename=".../figures/qRT-PCR/c_bcrp_trt_log_plot, height=4, w\\ \#ggsave(filename=".../figures/qRT-PCR/c_bcrp_trt_log_plot, height=4, w\\ \#ggsave(filename=".../figures/qRT-PCR/c_bcrp_trt_log_trt_log_trt_log_trt_log_trt_log_trt_log_trt_log_trt_log_trt_log_trt_log_trt_log_trt_log_trt_log_trt_log_trt_log_trt_lo$

Chronic morphine Bcrp test WITH P-gp in stats

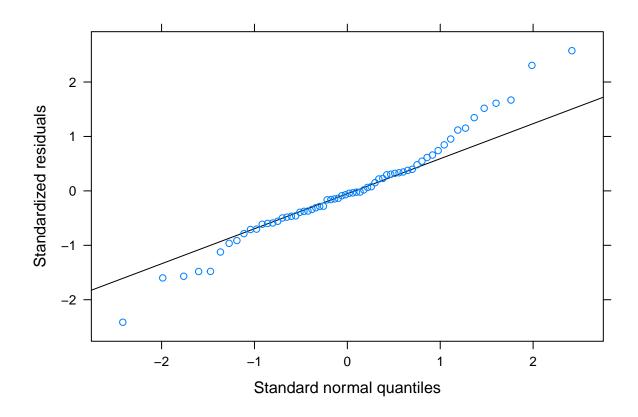
```
#bcrp & pgp together
chronic_long <- transporters_long %>% filter(Treatment %in% c("C", "S"))
chronic_long
```

```
## # A tibble: 64 x 5
     Sample Tissue Treatment gene
##
                                      rge
##
      <chr> <fct> <fct>
                              <fct> <dbl>
   1 C1
                   С
##
            hyp
                              Pgp
                                    1.77
                   С
##
   2 C1
            hyp
                              Bcrp
                                   1.62
                   С
   3 C1
                                    0.421
##
            ret
                             Pgp
##
   4 C1
            ret
                   C
                             Bcrp 0.892
  5 C2
                   С
                                    1.75
##
            hyp
                             Pgp
## 6 C2
                   С
                             Bcrp 1.63
            hyp
## 7 C2
                   С
                                   0.939
                             Pgp
            ret
```

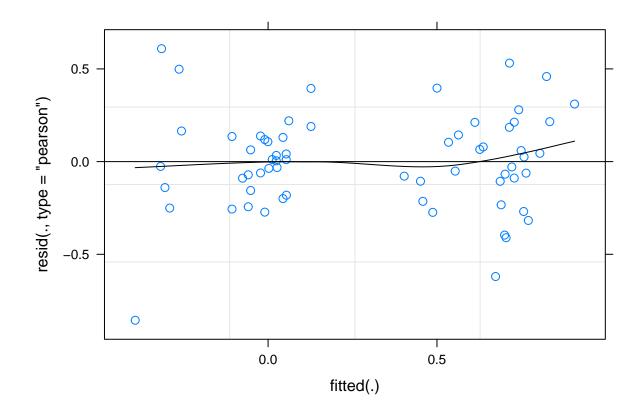
```
##
    8 C2
             ret
                    С
                               Bcrp 1.22
##
    9 C3
             hyp
                    С
                               Pgp
                                     1.23
                               Bcrp 2.37
## 10 C3
                    С
             hyp
## # ... with 54 more rows
chronic_lm <- lmer(rge ~ gene * Tissue * Treatment + (1|Sample), data=chronic_long)</pre>
plot(chronic_lm, type=c("p","smooth"), col.line=1)
```



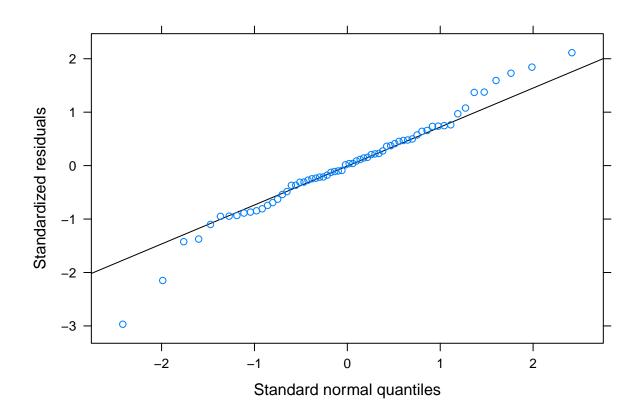
lattice::qqmath(chronic_lm)



```
chronic_log2_lm <- lmer(log2(rge) ~ gene * Tissue * Treatment + (1|Sample), data=chronic_long)
plot(chronic_log2_lm, type=c("p","smooth"), col.line=1)</pre>
```



lattice::qqmath(chronic_log2_lm)



```
shapiro.test(chronic_long$rge)
```

```
##
## Shapiro-Wilk normality test
##
## data: chronic_long$rge
## W = 0.95417, p-value = 0.01844
```

shapiro.test(log2(chronic_long\$rge))

```
##
## Shapiro-Wilk normality test
##
## data: log2(chronic_long$rge)
## W = 0.98001, p-value = 0.3853
```

anova(chronic_log2_lm)

```
## Type III Analysis of Variance Table with Satterthwaite's method
##
                         Sum Sq Mean Sq NumDF DenDF F value
                                                                Pr(>F)
## gene
                         0.0038 0.0038
                                            1
                                                 42
                                                      0.0453
                                                               0.83247
## Tissue
                         8.8468 8.8468
                                                 42 106.4695 4.365e-13 ***
                                            1
## Treatment
                         0.0190 0.0190
                                            1
                                                      0.2290
                                                               0.63967
                                                 42
                                                      3.8962
                                                               0.05500 .
## gene:Tissue
                         0.3237 0.3237
                                            1
```

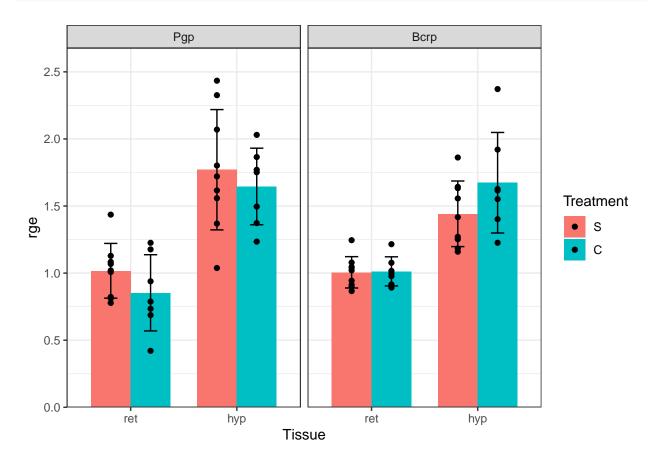
```
0.3575 0.3575 1
## gene:Treatment
                                             42
                                                  4.3023 0.04423 *
                                              42
## Tissue:Treatment
                      0.1717 0.1717
                                                  2.0665 0.15797
                                         1
## gene:Tissue:Treatment 0.0011 0.0011
                                              42 0.0127 0.91092
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
emmeans::emmeans(chronic_log2_lm, pairwise ~ Treatment | Tissue, by="gene")$contrasts
## Tissue = ret, gene = Pgp:
## contrast estimate SE
                          df t.ratio p.value
## S - C
            0.3061 0.154 53.9 1.982 0.0526
##
## Tissue = hyp, gene = Pgp:
## contrast estimate SE df t.ratio p.value
## S - C
           0.0809 0.154 53.9 0.524 0.6024
##
## Tissue = ret, gene = Bcrp:
## contrast estimate SE df t.ratio p.value
## S - C
          -0.0115 0.154 53.9 -0.075 0.9407
##
## Tissue = hyp, gene = Bcrp:
## contrast estimate
                       SE df t.ratio p.value
## S - C
           -0.2040 0.154 53.9 -1.321 0.1921
##
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
emmeans::emmeans(chronic_log2_lm, pairwise ~ Tissue | Treatment, by="gene")$contrasts
## Treatment = S, gene = Pgp:
## contrast estimate SE df t.ratio p.value
## ret - hyp -0.780 0.136 42 -5.742 <.0001
## Treatment = C, gene = Pgp:
## contrast estimate SE df t.ratio p.value
## ret - hyp -1.005 0.154 42 -6.525 <.0001
## Treatment = S, gene = Bcrp:
## contrast estimate SE df t.ratio p.value
## ret - hyp -0.510 0.136 42 -3.752 0.0005
##
## Treatment = C, gene = Bcrp:
## contrast estimate
                      SE df t.ratio p.value
## ret - hyp -0.702 0.154 42 -4.558 <.0001
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
emmeans::emmeans(chronic_log2_lm, pairwise ~ gene | Treatment, by="Tissue")$contrasts
## Treatment = S, Tissue = ret:
## contrast estimate
                        SE df t.ratio p.value
```

```
## Pgp - Bcrp 0.0000 0.136 42 0.000 1.0000
##
## Treatment = C, Tissue = ret:
## contrast estimate
                         SE df t.ratio p.value
## Pgp - Bcrp -0.3177 0.154 42 -2.062 0.0455
##
## Treatment = S, Tissue = hyp:
## contrast estimate
                          SE df t.ratio p.value
## Pgp - Bcrp 0.2704 0.136 42 1.990 0.0531
##
## Treatment = C, Tissue = hyp:
## contrast estimate
                          SE df t.ratio p.value
## Pgp - Bcrp -0.0146 0.154 42 -0.095 0.9251
##
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
chronic_sumstats <- chronic_long %>% group_by(Treatment, Tissue, gene) %>% summarise(
 n = n(),
 mean = mean(rge),
 sd = sd(rge),
 se = sd/sqrt(n),
 log2_mean = mean(log2(rge)),
 log2_sd = sd(log2(rge)),
 log2_se = log2_sd/sqrt(n)
)
## 'summarise()' has grouped output by 'Treatment', 'Tissue'. You can override
## using the '.groups' argument.
chronic sumstats
## # A tibble: 8 x 10
## # Groups: Treatment, Tissue [4]
                                                se log2_mean log2_sd log2_se
    Treatment Tissue gene
                              n mean
                                         sd
##
             <fct> <fct> <int> <dbl> <dbl> <dbl>
                                                     <dbl> <dbl> <dbl>
## 1 S
             ret
                              9 1.02 0.204 0.0680 -1.11e-11
                                                             0.280 0.0935
                     Pgp
## 2 S
             ret
                     Bcrp
                              9 1.01 0.117 0.0389 -4.39e-11 0.163 0.0542
## 3 S
                              9 1.77 0.448 0.149
                                                    7.80e- 1
                                                             0.385 0.128
             hyp
                     Pgp
                              9 1.44 0.244 0.0815 5.10e- 1
## 4 S
              hyp
                     Bcrp
                                                              0.242 0.0806
## 5 C
                              7 0.853 0.284 0.107 -3.06e- 1
              ret
                     Pgp
                                                              0.527 0.199
## 6 C
              ret
                     Bcrp
                              7 1.01 0.109 0.0411 1.15e- 2
                                                              0.150 0.0567
## 7 C
                              7 1.65 0.285 0.108
                                                    6.99e- 1
                                                               0.258 0.0976
              hyp
                     Pgp
                             7 1.67 0.375 0.142 7.14e- 1
## 8 C
                     Bcrp
                                                              0.308 0.116
              hyp
chronic_plot <- ggplot() +</pre>
 stat_summary(data=chronic_long, aes(x=Tissue, y=rge, fill=Treatment),
              fun = 'mean', geom="bar", width=0.75, position="dodge") +
 geom_errorbar(data=chronic_sumstats, aes(x=Tissue, ymin=mean-sd,
                                         ymax=mean+sd, fill=Treatment),
               width=0.25, position=position_dodge(width=0.75)) +
 geom_point(data=chronic_long, aes(x=Tissue, y=rge, fill=Treatment),
```

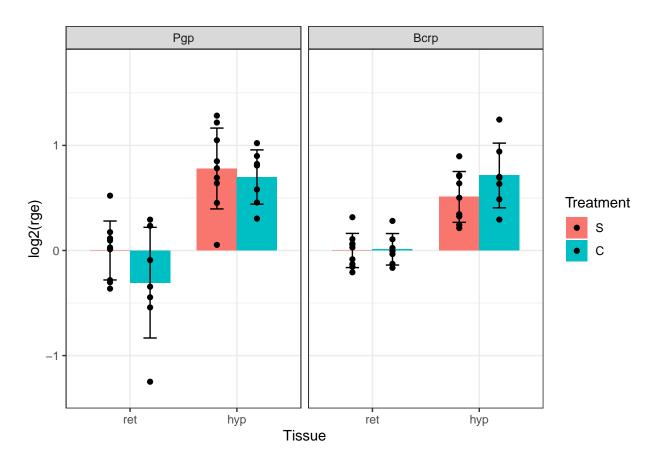
```
stat="identity", position=position_dodge(width=0.75)) +
facet_grid(~gene, scales="free_y") +
scale_y_continuous(expand = expansion(mult = c(0, 0.1))) +
#scale_fill_manual(values=c("maroon1", "springgreen3")) +
theme_bw()
```

Warning: Ignoring unknown aesthetics: fill

chronic_plot



chronic_log_plot



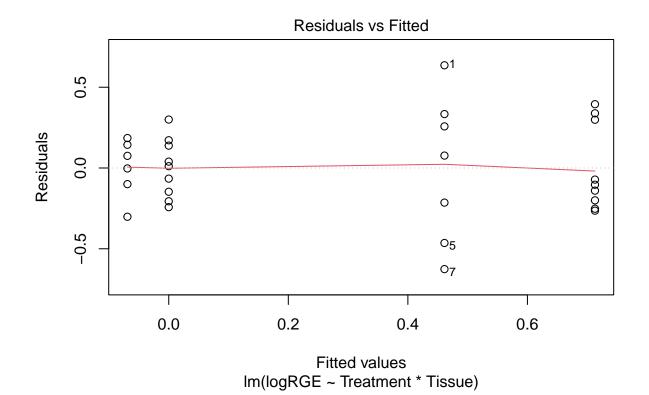
 $\#ggsave(filename=".../.../figures/qRT-PCR/chronic_trt_log.svg", ~plot=chronic_log_plot, ~height=4, ~width=6)$

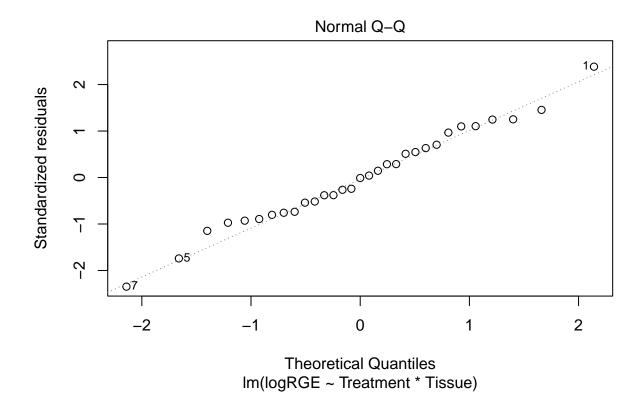
RERUN STATS WITH LOG2 ON DATA ORIGINALLY PUBLISHED IN BERGUM 2022 (p-gp chronic morphine/saline)

```
## Sample RelativeGE Tissue Treatment
## 1 M1 2.9955745 hypothalamus morphine
## 2 M2 2.2145575 hypothalamus morphine
## 3 M3 2.0521578 hypothalamus morphine
## 4 M4 1.7128570 hypothalamus morphine
## 5 M5 0.9969443 hypothalamus morphine
## 6 M6 1.2801494 hypothalamus morphine
```

```
## 7
          M7 0.8480968 hypothalamus morphine
## 8
          S1 1.7732013 hypothalamus
                                        saline
## 9
          S2 1.8987767 hypothalamus
                                        saline
## 10
          S3 1.8391614 hypothalamus
                                        saline
## 11
          S4 1.5664282 hypothalamus
                                        saline
## 12
          S5 1.5853405 hypothalamus
                                        saline
## 13
          S6 1.6699513 hypothalamus
                                        saline
## 14
          S7 3.0282725 hypothalamus
                                        saline
## 15
          S8 2.7501401 hypothalamus
                                        saline
## 16
          S9 2.8621095 hypothalamus
                                        saline
## 17
          M2 1.0775971
                              retina
                                     morphine
## 18
          M3 0.9306124
                                      morphine
                              retina
## 19
          M4 1.1232327
                              retina
                                      morphine
## 20
                                      morphine
          M5 1.0061231
                              retina
## 21
          M6 0.6900942
                              retina
                                      morphine
## 22
          M7 0.8444983
                              retina
                                      morphine
## 23
          S1 0.8139308
                                        saline
                              retina
## 24
          S2 1.0113566
                              retina
                                        saline
## 25
          S3 1.0402547
                              retina
                                        saline
          S4 1.1486002
## 26
                              retina
                                        saline
## 27
          S5 0.8632570
                              retina
                                        saline
## 28
          S6 0.9363421
                              retina
                                        saline
## 29
          S7 1.3498949
                                        saline
                              retina
## 30
          S8
              0.7846937
                              retina
                                        saline
## 31
          S9 1.1874759
                              retina
                                        saline
shapiro.test(pgp$RelativeGE)
##
   Shapiro-Wilk normality test
##
## data: pgp$RelativeGE
## W = 0.86255, p-value = 0.0009488
shapiro.test(sqrt(pgp$RelativeGE))
##
##
   Shapiro-Wilk normality test
##
## data: sqrt(pgp$RelativeGE)
## W = 0.90574, p-value = 0.01005
shapiro.test(log(pgp$RelativeGE))
##
##
   Shapiro-Wilk normality test
##
## data: log(pgp$RelativeGE)
## W = 0.93745, p-value = 0.07015
```

```
shapiro.test(log2(pgp$RelativeGE))
##
##
   Shapiro-Wilk normality test
## data: log2(pgp$RelativeGE)
## W = 0.93745, p-value = 0.07015
pgp <- pgp %>% mutate(logRGE = log(RelativeGE))
pgp <- pgp %>% mutate(log2RGE = log2(RelativeGE))
car::leveneTest(logRGE ~ Tissue * Treatment, pgp)
## Levene's Test for Homogeneity of Variance (center = median)
         Df F value Pr(>F)
## group 3 2.6055 0.0723 .
##
         27
## -
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
pgp_log_lm <- lm(logRGE ~ Treatment * Tissue, pgp)</pre>
plot(pgp_log_lm, which=c(1,2))
```

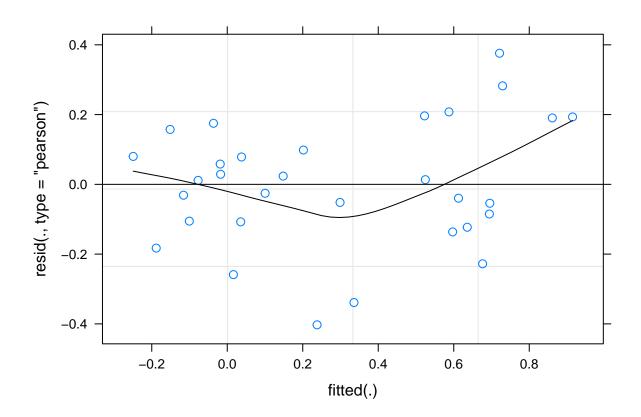




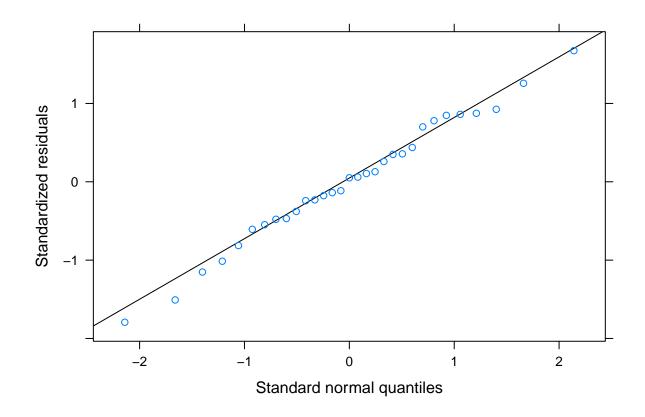
anova(pgp_log_lm)

emmeans::emmeans(pgp_log_lm, pairwise ~ Tissue | Treatment)\$contrasts

```
## Treatment = saline:
##
    contrast
                          estimate
                                      SE df t.ratio p.value
                            -0.713 0.136 27 -5.251 <.0001
    retina - hypothalamus
##
## Treatment = morphine:
   contrast
                                      SE df t.ratio p.value
##
                          estimate
   retina - hypothalamus
                           -0.530 0.160 27 -3.310 0.0027
pgp_log_lmer <- lmer(log(RelativeGE) ~ Tissue * Treatment + (1|Sample), data =pgp)</pre>
plot(pgp_log_lmer, type=c("p","smooth"), col.line=1)
```



lattice::qqmath(pgp_log_lmer)

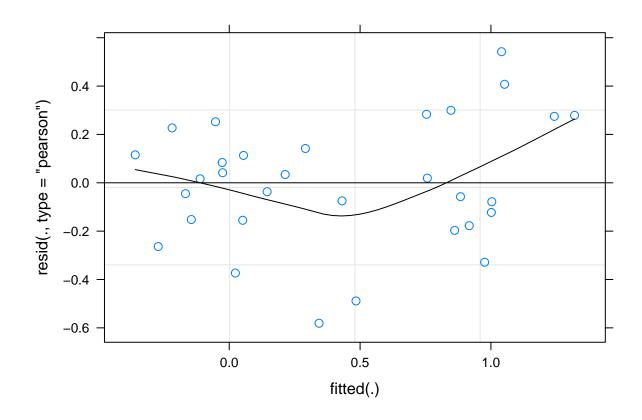


anova(pgp_log_lmer)

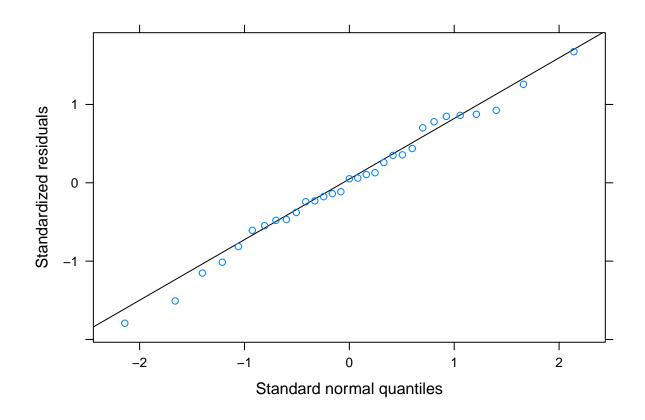
```
## Type III Analysis of Variance Table with Satterthwaite's method
                     Sum Sq Mean Sq NumDF DenDF F value
                    2.65924 2.65924
## Tissue
                                        1 11.304 52.7214 1.395e-05 ***
## Treatment
                    0.06191 0.06191
                                        1 11.739
                                                 1.2274
                                                            0.2901
## Tissue:Treatment 0.09433 0.09433
                                        1 11.304
                                                 1.8701
                                                            0.1980
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
emmeans::emmeans(pgp_log_lmer, pairwise ~ Treatment | Tissue)
```

```
## $emmeans
## Tissue = retina:
   Treatment emmean
                          SE
                               df lower.CL upper.CL
               0.0000 0.0974 23.3
                                     -0.201
   saline
                                               0.201
##
   morphine -0.0258 0.1187 24.8
                                    -0.270
                                               0.219
##
## Tissue = hypothalamus:
  Treatment emmean
                          SE
                               df lower.CL upper.CL
   saline
               0.7130 0.0974 23.3
                                     0.512
                                               0.914
               0.4612 0.1104 23.3
                                     0.233
                                               0.689
##
   morphine
## Degrees-of-freedom method: kenward-roger
```

```
## Results are given on the log (not the response) scale.
## Confidence level used: 0.95
##
## $contrasts
## Tissue = retina:
## contrast estimate
                                SE df t.ratio p.value
## saline - morphine 0.0258 0.154 24.2 0.168 0.8680
##
## Tissue = hypothalamus:
## contrast
                     estimate
                                SE
                                     df t.ratio p.value
                                         1.710 0.1005
## saline - morphine
                     0.2518 0.147 23.3
## Degrees-of-freedom method: kenward-roger
## Results are given on the log (not the response) scale.
emmeans::emmeans(pgp_log_lmer, pairwise ~ Tissue | Treatment)
## $emmeans
## Treatment = saline:
              emmean
## Tissue
                           SE df lower.CL upper.CL
## retina
                0.0000 0.0974 23.3 -0.201
                                              0.201
## hypothalamus 0.7130 0.0974 23.3
                                      0.512
                                               0.914
##
## Treatment = morphine:
## Tissue
                               df lower.CL upper.CL
                emmean
                           SE
## retina
                -0.0258 0.1187 24.8 -0.270
                                              0.219
## hypothalamus 0.4612 0.1104 23.3
                                      0.233
                                               0.689
## Degrees-of-freedom method: kenward-roger
## Results are given on the log (not the response) scale.
## Confidence level used: 0.95
##
## $contrasts
## Treatment = saline:
                                    SE df t.ratio p.value
## contrast
                        estimate
## retina - hypothalamus -0.713 0.106 13.1 -6.734 <.0001
## Treatment = morphine:
                         estimate
                                    SE df t.ratio p.value
## contrast
## retina - hypothalamus -0.487 0.128 13.9 -3.812 0.0019
## Degrees-of-freedom method: kenward-roger
## Results are given on the log (not the response) scale.
pgp_log2_lmer <- lmer(log2(RelativeGE) ~ Tissue * Treatment + (1|Sample), data =pgp)
plot(pgp_log2_lmer, type=c("p","smooth"), col.line=1)
```



lattice::qqmath(pgp_log2_lmer)



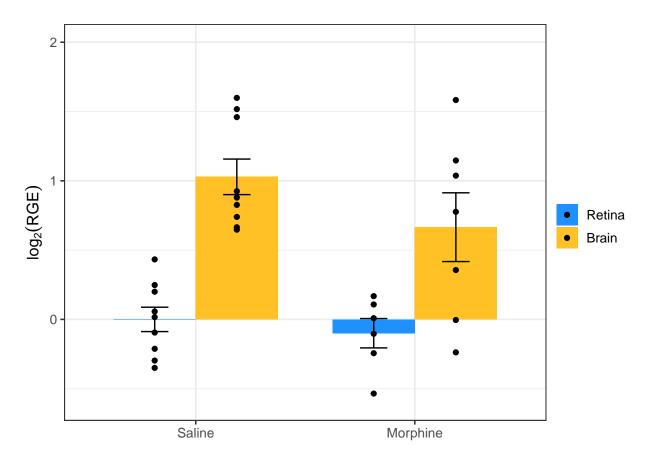
anova(pgp_log2_lmer)

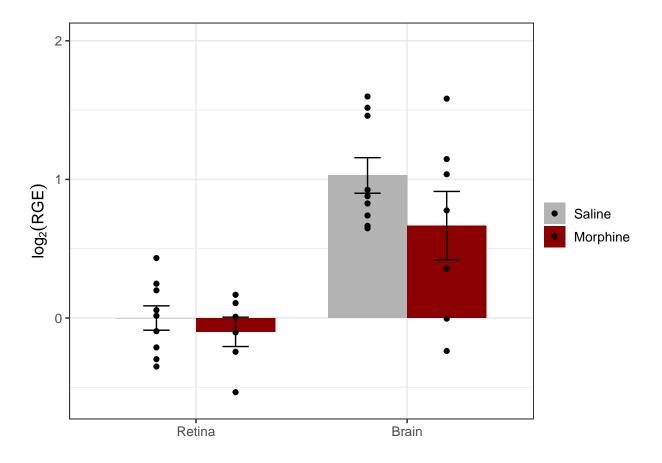
```
## Type III Analysis of Variance Table with Satterthwaite's method
                   Sum Sq Mean Sq NumDF DenDF F value
                   5.5349 5.5349
## Tissue
                                      1 11.304 52.7214 1.395e-05 ***
## Treatment
                   0.1289 0.1289
                                      1 11.739
                                               1.2274
                                                         0.2901
## Tissue:Treatment 0.1963 0.1963
                                      1 11.304 1.8701
                                                         0.1980
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
emmeans::emmeans(pgp_log2_lmer, pairwise ~ Treatment | Tissue)
```

```
## $emmeans
## Tissue = retina:
   Treatment emmean
                         SE
                              df lower.CL upper.CL
               0.0000 0.140 23.3
                                   -0.290
                                              0.290
   saline
##
   morphine -0.0372 0.171 24.8
                                   -0.390
                                              0.316
##
## Tissue = hypothalamus:
   Treatment emmean
                         SE
                              df lower.CL upper.CL
   saline
               1.0286 0.140 23.3
                                     0.738
                                              1.319
                                              0.995
               0.6654 0.159 23.3
                                     0.336
##
   morphine
```

Degrees-of-freedom method: kenward-roger

```
## Results are given on the log2 (not the response) scale.
## Confidence level used: 0.95
##
## $contrasts
## Tissue = retina:
## contrast
                                 SE df t.ratio p.value
                     estimate
## saline - morphine 0.0372 0.222 24.2
                                         0.168 0.8680
##
## Tissue = hypothalamus:
## contrast
                     estimate
                                 SE
                                      df t.ratio p.value
## saline - morphine
                     0.3633 0.212 23.3
                                         1.710 0.1005
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
emmeans::emmeans(pgp_log2_lmer, pairwise ~ Tissue | Treatment)
## $emmeans
## Treatment = saline:
## Tissue
                           SE
                              df lower.CL upper.CL
                emmean
## retina
                 0.0000 0.140 23.3
                                   -0.290
## hypothalamus 1.0286 0.140 23.3
                                      0.738
                                               1.319
##
## Treatment = morphine:
## Tissue
                emmean
                           SE
                                df lower.CL upper.CL
## retina
                -0.0372 0.171 24.8
                                   -0.390
## hypothalamus 0.6654 0.159 23.3
                                      0.336
                                              0.995
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
## Confidence level used: 0.95
## $contrasts
## Treatment = saline:
                                    SE df t.ratio p.value
## contrast
                         estimate
## retina - hypothalamus -1.029 0.153 13.1 -6.734 <.0001
##
## Treatment = morphine:
## contrast
                         estimate
                                     SE df t.ratio p.value
## retina - hypothalamus -0.703 0.184 13.9 -3.812 0.0019
##
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
pgp_log2_tiss_plot <- pgp %>% ggplot(aes(x=Treatment, y=log2RGE, fill=Tissue)) +
 stat_summary(fun = 'mean', geom="bar", aes(fill=Tissue),
              width=0.75, position="dodge") +
 stat_summary(fun.data = mean_se, geom = "errorbar", width=0.25,
              position=position_dodge(width=0.75)) +
 geom_point(stat="identity", position=position_dodge(width=0.75)) +
 theme_bw(base_size = 12) %+replace%
 theme(axis.title.x=element blank(),
       legend.title = element_blank(),
```





 $\#ggsave(filename = "../../figures/pgp_log2_trt_plot.png", plot = pgp_log2_trt_plot, height=4, width=4)$

Looking at only the 1 injection samples (males, low e/p females, high e/p females)

Reading in the new dataset

```
transporters <- read.csv(".../.../data/qRT-PCR/07222022-transporter-gene-study-linj-only.csv", fileEncodic
#transporters <- transporters %>% filter(Sample != "calib")
```

```
transporters <- transporters %>% mutate(Tissue = str_sub(Sample, start=-3L, end=-1L),
                                       Group = str_extract(Sample, "[:alpha:]+(?=[:digit:])"),
                                       Sample = str_extract(Sample, "[:alnum:]+(?=-)"))
transporters <- transporters %>% mutate(Group = factor(Group, c("M", "Flo", "Fhi")),
                                       Tissue = factor(Tissue, c("ret", "hyp")),
                                       Sample = as.factor(Sample))
transporters_long <- transporters %>% pivot_longer(cols=c("Pgp.RGE", "Bcrp.RGE"), names_to = "gene", va
transporters_long <- transporters_long %>% mutate(gene = str_extract(gene, "[:alpha:]+(?=.)"))
transporters_long <- transporters_long %>% mutate(gene = factor(as.factor(gene), c("Pgp", "Bcrp")))
transporters_long
## # A tibble: 96 x 5
     Sample Tissue Group gene
##
     <fct> <fct> <fct> <fct> <fct> <dbl>
## 1 Fhi10 hyp
                   Fhi
                         Pgp
                               4.38
## 2 Fhi10 hyp
                         Bcrp 1.50
                   Fhi
## 3 Fhi10 ret
                   Fhi
                               1.22
                         Pgp
## 4 Fhi10 ret
                   Fhi
                         Bcrp 0.905
## 5 Fhi11 hyp
                 Fhi
                         Pgp 3.29
## 6 Fhi11 hyp
                   Fhi
                         Bcrp 1.71
## 7 Fhi11 ret
                   Fhi
                         Pgp
                               1.08
## 8 Fhi11 ret
                   Fhi
                         Bcrp 0.880
```

Stats

9 Fhi1 hyp

hyp ## # ... with 86 more rows

10 Fhi1

Fhi

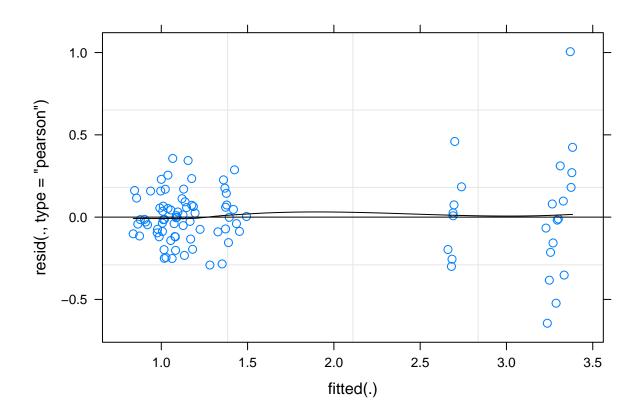
Fhi

Pgp

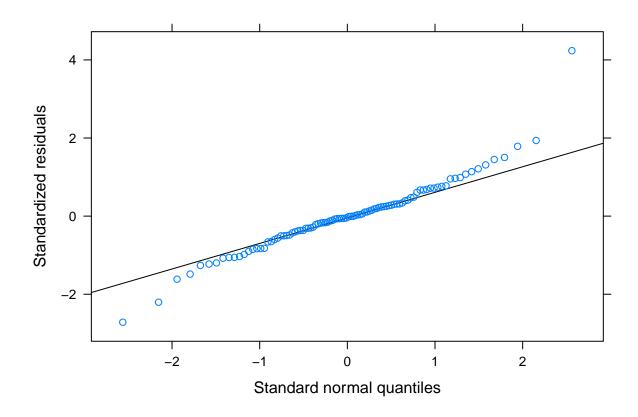
Bcrp 1.39

3.11

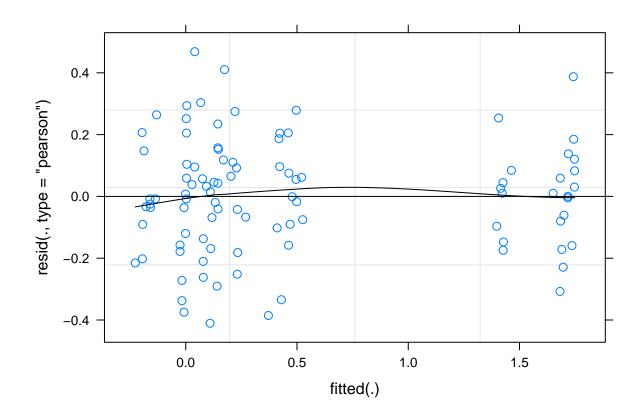
```
transporter_lm <- lmer(rge ~ gene * Tissue * Group + (1|Sample), data=transporters_long)
plot(transporter_lm, type=c("p","smooth"), col.line=1)
```



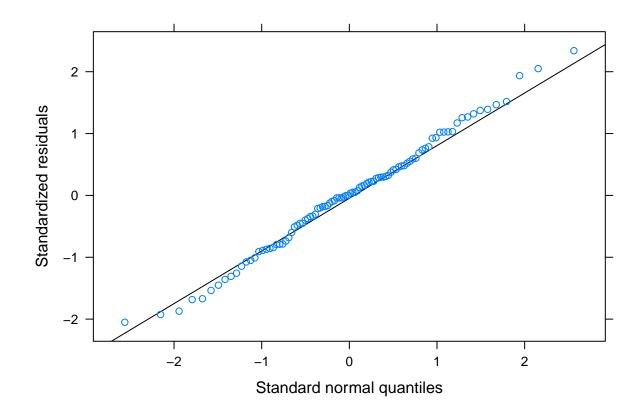
lattice::qqmath(transporter_lm)



```
 transporter_log2_lm <- lmer(log2(rge) ~ gene * Tissue * Group + (1|Sample), \\ \frac{data=}{data=} transporters_long) \\ plot(transporter_log2_lm, \\ \frac{type=}{data=} transporter_long) \\ plot(transporter_long2_lm, \\ \frac{type=}{data=} transporter_long) \\ plot(transporter_long2_lm, \\ \frac{type=}{data=} transporter_long) \\ plot(transporter_long2_lm, \\ \frac{type=}{data=} transporter_long2_lm, \\ \frac{type=}{data=}
```



lattice::qqmath(transporter_log2_lm)



```
shapiro.test(transporters_long$rge)

##
## Shapiro-Wilk normality test
##
## data: transporters_long$rge
## W = 0.77877, p-value = 1.044e-10

shapiro.test(log2(transporters_long$rge))

##
## Shapiro-Wilk normality test
##
## data: log2(transporters_long$rge)
## W = 0.87148, p-value = 1.285e-07
anova(transporter_log2_lm)
```

Sum Sq Mean Sq NumDF DenDF F value Pr(>F)

63 253.4878 < 2e-16 ***

63 509.4664 < 2e-16 ***

63 170.2831 < 2e-16 ***

4.2854 0.02749 *

1

1

2

Type III Analysis of Variance Table with Satterthwaite's method

10.1573 10.1573

20.4143 20.4143

0.3434 0.1717

6.8233 6.8233

##

gene

Tissue

gene:Tissue

Group

```
## gene:Group
                    0.1278 0.0639
                                      2
                                           63
                                                1.5945 0.21109
                    0.3872 0.1936
                                      2
                                           63 4.8315 0.01117 *
## Tissue:Group
                                           63 1.7536 0.18149
## gene:Tissue:Group 0.1405 0.0703
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
emmeans::emmeans(transporter_log2_lm, pairwise ~ Group | Tissue, by="gene")$contrasts
## Tissue = ret, gene = Pgp:
## contrast estimate
                         SE df t.ratio p.value
## M - Flo
             -0.1312 0.1180 83 -1.111 0.5098
## M - Fhi
             -0.1306 0.0962 83 -1.357 0.3680
## Flo - Fhi 0.0006 0.1117 83 0.005 1.0000
## Tissue = hyp, gene = Pgp:
## contrast estimate
                         SE df t.ratio p.value
## M - Flo -0.3108 0.1180 83 -2.633 0.0270
## M - Fhi
            -0.2798 0.0962 83 -2.909 0.0128
## Flo - Fhi 0.0310 0.1117 83 0.277 0.9585
## Tissue = ret, gene = Bcrp:
## contrast estimate
                         SE df t.ratio p.value
## M - Flo -0.0658 0.1180 83 -0.557 0.8430
## M - Fhi 0.1748 0.0962 83 1.818 0.1702
## Flo - Fhi 0.2406 0.1117 83 2.155 0.0852
##
## Tissue = hyp, gene = Bcrp:
## contrast estimate
                        SE df t.ratio p.value
           -0.1804 0.1180 83 -1.528 0.2830
## M - Flo
## M - Fhi
             -0.2530 0.0962 83 -2.630 0.0272
## Flo - Fhi -0.0726 0.1117 83 -0.651 0.7926
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
## P value adjustment: tukey method for comparing a family of 3 estimates
emmeans::emmeans(transporter_log2_lm, pairwise ~ Tissue | Group, by="gene")$contrasts
## Group = M, gene = Pgp:
## contrast estimate
                         SE df t.ratio p.value
## ret - hyp -1.423 0.1001 63 -14.219 <.0001
##
## Group = Flo, gene = Pgp:
## contrast estimate SE df t.ratio p.value
## ret - hyp -1.603 0.1266 63 -12.660 <.0001
##
## Group = Fhi, gene = Pgp:
## contrast estimate
                         SE df t.ratio p.value
## ret - hyp -1.572 0.0854 63 -18.422 <.0001
## Group = M, gene = Bcrp:
## contrast estimate
                         SE df t.ratio p.value
```

ret - hyp -0.229 0.1001 63 -2.287 0.0255

```
##
## Group = Flo, gene = Bcrp:
## contrast estimate
                        SE df t.ratio p.value
## ret - hyp -0.344 0.1266 63 -2.714 0.0086
## Group = Fhi, gene = Bcrp:
## contrast estimate
                         SE df t.ratio p.value
## ret - hyp -0.657 0.0854 63 -7.695 <.0001
##
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
emmeans::emmeans(transporter_log2_lm, pairwise ~ gene | Group, by="Tissue")$contrasts
## Group = M, Tissue = ret:
## contrast estimate
                           SE df t.ratio p.value
## Pgp - Bcrp 0.0000 0.1001 63 0.000 1.0000
##
## Group = Flo, Tissue = ret:
## contrast estimate
                          SE df t.ratio p.value
## Pgp - Bcrp 0.0654 0.1266 63
                                 0.516 0.6073
##
## Group = Fhi, Tissue = ret:
## contrast estimate
                         SE df t.ratio p.value
## Pgp - Bcrp 0.3054 0.0854 63
                                  3.578 0.0007
##
## Group = M, Tissue = hyp:
## contrast estimate
                          SE df t.ratio p.value
## Pgp - Bcrp 1.1942 0.1001 63 11.932 <.0001
##
## Group = Flo, Tissue = hyp:
## contrast estimate
                          SE df t.ratio p.value
## Pgp - Bcrp 1.3247 0.1266 63 10.463 <.0001
##
## Group = Fhi, Tissue = hyp:
## contrast estimate
                         SE df t.ratio p.value
## Pgp - Bcrp 1.2211 0.0854 63 14.306 <.0001
##
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
trans_sumstats <- transporters_long %>% group_by(Group, Tissue, gene) %>% summarise(
 n = n(),
 mean = mean(rge),
 sd = sd(rge),
 se = sd/sqrt(n),
 log2_mean = mean(log2(rge)),
 log2_sd = sd(log2(rge)),
 log2_se = log2_sd/sqrt(n)
)
```

'summarise()' has grouped output by 'Group', 'Tissue'. You can override using

the '.groups' argument.

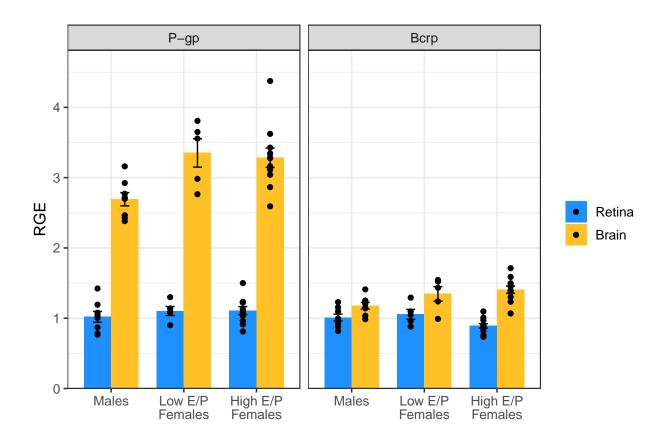
```
trans sumstats
```

```
## # A tibble: 12 x 10
## # Groups: Group, Tissue [6]
##
     Group Tissue gene
                           n mean
                                      sd
                                            se log2_mean log2_sd log2_se
##
     <fct> <fct> <fct> <int> <dbl> <dbl> <dbl>
                                                   <dbl>
                                                           <dbl>
                                                                  <dbl>
##
  1 M
                           8 1.02 0.220 0.0779 2.02e-10
                                                           0.305 0.108
           ret
                 Pgp
                           8 1.01 0.142 0.0501 6.58e-11
## 2 M
           ret
                 Bcrp
                                                           0.202 0.0713
## 3 M
                           8 2.69 0.266 0.0942 1.42e+ 0
                                                          0.141 0.0498
           hyp
                 Pgp
## 4 M
           hyp
                  Bcrp
                           8 1.18 0.133 0.0470 2.29e- 1
                                                           0.163 0.0576
## 5 Flo
                           5 1.10 0.143 0.0638 1.31e- 1
                                                           0.190 0.0848
           ret
                 Pgp
## 6 Flo
           ret
                           5 1.06 0.156 0.0699 6.58e- 2
                                                           0.209 0.0935
                 Bcrp
                           5 3.35 0.452 0.202
## 7 Flo
                 Pgp
                                                1.73e+ 0
                                                           0.201 0.0897
           hyp
## 8 Flo
           hyp
                 Bcrp
                           5 1.35 0.232 0.104
                                                4.09e- 1
                                                           0.268 0.120
## 9 Fhi
                          11 1.11 0.187 0.0565 1.31e- 1
                                                           0.242 0.0731
           ret
                 Pgp
                          11 0.891 0.105 0.0316 -1.75e- 1
## 10 Fhi
           ret
                 Bcrp
                                                           0.168 0.0507
## 11 Fhi
                          11 3.28 0.457 0.138
                                                1.70e+ 0
                                                           0.193 0.0580
           hyp
                  Pgp
## 12 Fhi
                          11 1.41 0.172 0.0519 4.82e- 1
                                                           0.182 0.0548
           hyp
                  Bcrp
```

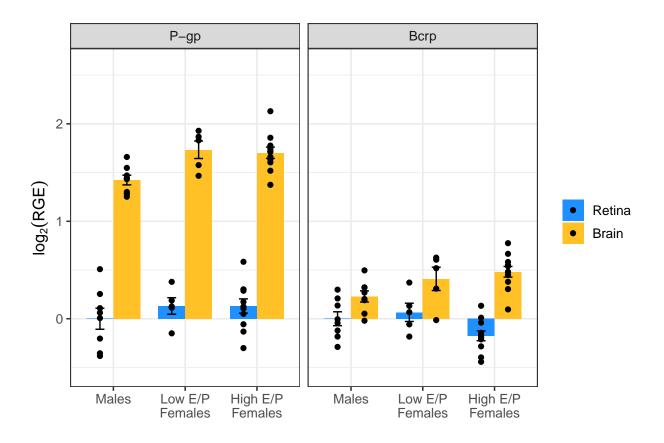
Visualizing hyp vs ret within each treatment/gene

```
gene.labs <- c("P-gp", "Bcrp")</pre>
names(gene.labs) <- c("Pgp", "Bcrp")</pre>
trans_hyp_ret_plot <- ggplot() +</pre>
  stat_summary(data=transporters_long, aes(x=Group, y=rge, fill=Tissue),
               fun = 'mean', geom="bar", width=0.75, position="dodge") +
  geom_errorbar(data=trans_sumstats, aes(x=Group, ymin=mean-se, ymax=mean+se, fill=Tissue),
                width=0.25, position=position_dodge(width=0.75)) +
  geom_point(data=transporters_long, aes(x=Group, y=rge, fill=Tissue),
             stat="identity", position=position_dodge(width=0.75)) +
  facet_wrap(~gene, labeller = labeller(gene = gene.labs), ncol=2) +
  scale_y = continuous(expand = expansion(mult = c(0, 0.1))) +
  theme bw(base size = 12) %+replace%
  theme(legend.title = element blank()) +
  scale_fill_manual(values=c("dodgerblue", "goldenrod1"),
                    labels=c("Retina", "Brain")) +
  scale_x_discrete(labels=c("M" = "Males", "Flo" = "Low E/P\nFemales",
                             "Fhi" = "High E/P \in (N - 1)" +
  xlab("") +
  ylab("RGE")
```

```
trans_hyp_ret_plot
```



```
\#ggsave(filename="../../figures/qRT-PCR/trans\_hyp\_ret\_1inj.png",\ plot=trans\_hyp\_ret\_plot,\ height=4,\ widship in the property of the proper
\#ggsave(filename="../../figures/qRT-PCR/trans\_hyp\_ret\_1inj.svg",\ plot=trans\_hyp\_ret\_plot,\ height=4,\ widship in the property of the proper
trans_hyp_ret_log_plot <- ggplot() +</pre>
         stat_summary(data=transporters_long, aes(x=Group, y=log2(rge), fill=Tissue),
                                                                    fun = 'mean', geom="bar", width=0.75, position="dodge") +
         geom_errorbar(data=trans_sumstats, aes(x=Group, ymin=log2_mean-log2_se,
                                                                                                                                                                                           ymax=log2_mean+log2_se, fill=Tissue),
                                                                         width=0.25, position=position_dodge(width=0.75)) +
         geom_point(data=transporters_long, aes(x=Group, y=log2(rge), fill=Tissue),
                                                           stat="identity", position=position_dodge(width=0.75)) +
        facet_wrap(~gene, labeller = labeller(gene = gene.labs), ncol=2) +
         scale_y_continuous(expand = expansion(mult = c(0.1, 0.25))) +
         theme_bw(base_size = 12) %+replace%
         theme(legend.title = element_blank()) +
         scale_fill_manual(values=c("dodgerblue", "goldenrod1"),
                                                                                           labels=c("Retina", "Brain")) +
         scale_x_discrete(labels=c("M" = "Males", "Flo" = "Low E/P\nFemales",
                                                                                                                                 "Fhi" = "High E/P\nFemales")) +
         xlab("") +
         ylab(expression(log[2](RGE)))
```



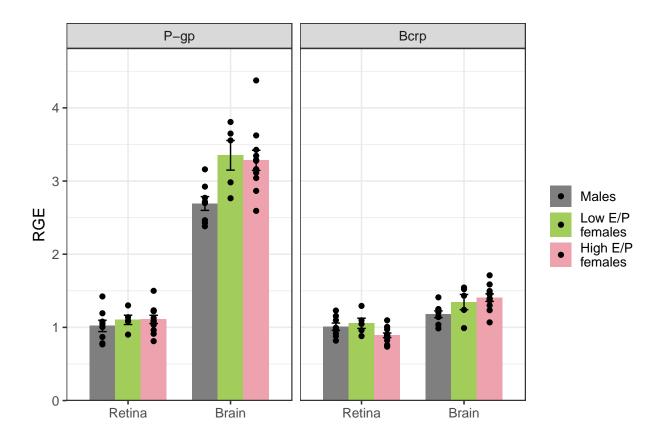
 $\#ggsave(filename="../../figures/qRT-PCR/trans_hyp_ret_log_1inj.png", plot=trans_hyp_ret_log_plot, heightfollowed filename="../../figures/qRT-PCR/trans_hyp_ret_log_1inj.svg", plot=trans_hyp_ret_log_plot, heightfollowed filename="../../figures/qRT-PCR/trans_hyp_ret_log_1inj.svg", plot=trans_hyp_ret_log_plot, heightfollowed filename=".../.../figures/qRT-PCR/trans_hyp_ret_log_1inj.svg", plot=trans_hyp_ret_log_1inj.svg", plot=trans_hyp_1inj.svg", plot=trans_hyp_1inj.svg", plot=trans_hyp_1inj.svg", plot=trans_hyp_1inj.svg", plot=trans_hyp_1inj.svg$

Visualizing sex diffs within each tissue/gene

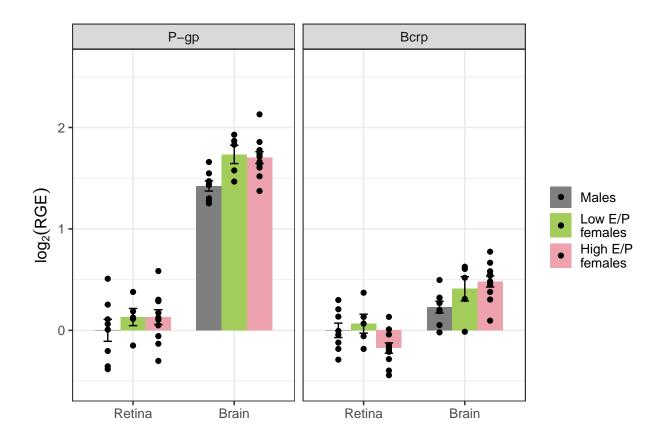
```
trans_trt_plot <- ggplot() +</pre>
  stat_summary(data=transporters_long, aes(x=Tissue, y=rge, fill=Group),
               fun = 'mean', geom="bar", width=0.75, position="dodge") +
  geom_errorbar(data=trans_sumstats, aes(x=Tissue, ymin=mean-se, ymax=mean+se, fill=Group),
                width=0.25, position=position_dodge(width=0.75)) +
  geom_point(data=transporters_long, aes(x=Tissue, y=rge, fill=Group),
             stat="identity", position=position_dodge(width=0.75)) +
  facet_wrap(~gene, labeller = labeller(gene = gene.labs), ncol=2) +
  scale_y_continuous(expand = expansion(mult = c(0, 0.1))) +
  theme_bw(base_size = 12) %+replace%
  theme(legend.title = element_blank(),
        legend.spacing.y = unit(0.1, 'cm')) +
  ## next line needed to make previous line work
  guides(fill = guide_legend(byrow = TRUE)) +
  scale_fill_manual(values=c("gray50", "darkolivegreen3", "lightpink2"),
                    labels=c("M" = "Males", "Flo" = "Low E/P\nfemales",
```

```
"Fhi" = "High E/P\nfemales")) +
scale_x_discrete(labels=c("ret" = "Retina", "hyp" = "Brain")) +
xlab("") +
ylab("RGE")
```

trans_trt_plot



trans_trt_log_plot



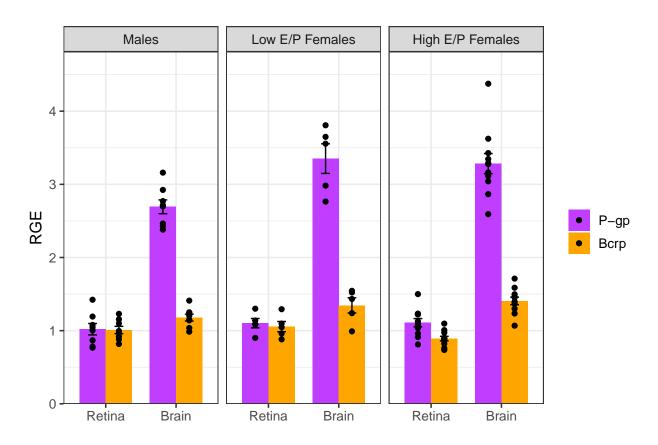
 $\#ggsave(filename=".../.../figures/qRT-PCR/trans_trt_log_1inj.png", plot=trans_trt_log_plot, height=4, widfiggsave(filename=".../.../figures/qRT-PCR/trans_trt_log_1inj.svg", plot=trans_trt_log_plot, height=4, widfigures/qRT-PCR/trans_trt_log_1inj.svg", plot=trans_trt_log_plot, height=4, widfigures/qRT-PCR/trans_trt_log_plot, height=4, widfigures/qRT-PCR/trans_trt$

Visualizing P-gp vs Bcrp differences

```
group.labs <- c("Males", "Low E/P Females", "High E/P Females")
names(group.labs) <- c("M", "Flo", "Fhi")</pre>
```

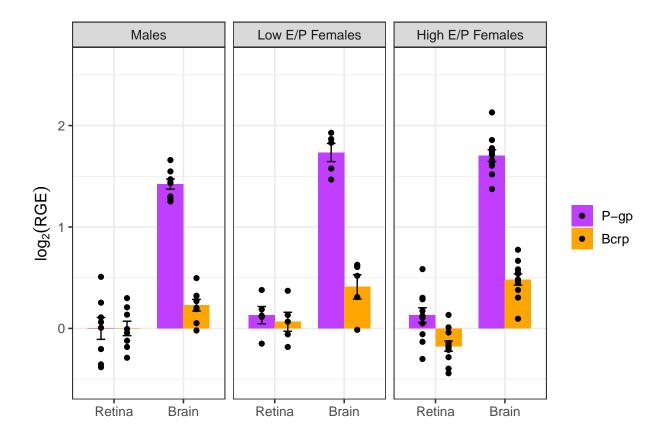
```
trans_g2g_plot <- ggplot() +</pre>
  stat_summary(data=transporters_long, aes(x=Tissue, y=rge, fill=gene),
               fun = 'mean', geom="bar", width=0.75, position="dodge") +
  geom_errorbar(data=trans_sumstats, aes(x=Tissue, ymin=mean-se,
                                         ymax=mean+se, fill=gene),
                width=0.25, position=position_dodge(width=0.75)) +
  geom_point(data=transporters_long, aes(x=Tissue, y=rge, fill=gene),
             position=position dodge(width=0.75)) +
  facet_grid(~Group, labeller = labeller(Group = group.labs)) +
  scale_y_continuous(expand = expansion(mult = c(0, 0.1))) +
  scale_fill_manual(values=c("darkorchid1", "orange1"),
                    labels = c("P-gp", "Bcrp")) +
  theme_bw(base_size = 12) %+replace%
  theme(legend.title = element_blank()) +
  xlab("") +
  ylab("RGE") +
  scale_x_discrete(labels=c("ret" = "Retina", "hyp" = "Brain"))
```

${\tt trans_g2g_plot}$



```
\#ggsave(filename=".../../figures/qRT-PCR/trans_g2g_1inj.png", plot=trans_g2g_plot, height=4, width=7)
#qqsave(filename="../../fiqures/qRT-PCR/trans_q2q_1inj.svq", plot=trans_q2q_plot, height=4, width=7)
trans_g2g_log_plot <- ggplot() +</pre>
  stat_summary(data=transporters_long, aes(x=Tissue, y=log2(rge), fill=gene),
               fun = 'mean', geom="bar", width=0.75, position="dodge") +
  geom errorbar(data=trans sumstats, aes(x=Tissue, ymin=log2 mean-log2 se,
                                         ymax=log2_mean+log2_se, fill=gene),
                width=0.25, position=position_dodge(width=0.75)) +
  geom_point(data=transporters_long, aes(x=Tissue, y=log2(rge), fill=gene),
             position=position_dodge(width=0.75)) +
  facet wrap(~Group, labeller = labeller(Group = group.labs)) +
  scale_y_continuous(expand = expansion(mult = c(0.1, 0.25))) +
  scale_fill_manual(values=c("darkorchid1", "orange1"),
                    labels = c("P-gp", "Bcrp")) +
  theme_bw(base_size = 12) %+replace%
  theme(legend.title = element_blank()) +
  xlab("") +
  ylab(expression(log[2](RGE))) +
  scale_x_discrete(labels=c("ret" = "Retina", "hyp" = "Brain"))
```

trans_g2g_log_plot

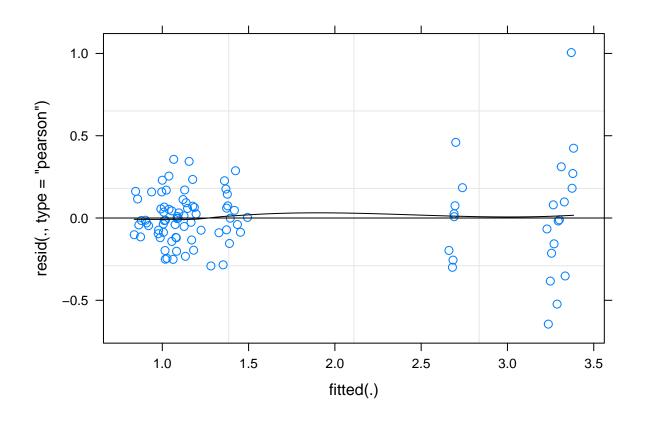


```
\#ggsave(filename=".../sigures/qRT-PCR/trans\_g2g\_log\_1inj.png", plot=trans\_g2g\_log\_plot, height=4, wid\#ggsave(filename=".../sigures/qRT-PCR/trans\_g2g\_log\_1inj.svg", plot=trans\_g2g\_log\_plot, height=4, wid#ggsave(filename=".../sigures/qRT-PCR/trans\_g2g\_log\_1inj.svg", plot=trans\_g2g\_log\_plot, height=4, wid#ggsave(filename=".../sigures/qRT-PCR/trans\_g2g\_log\_plot, height=4, wid#gasyave(filename=".../sigures/qRT-PCR/trans\_g2g\_log\_plot, height=4, w
```

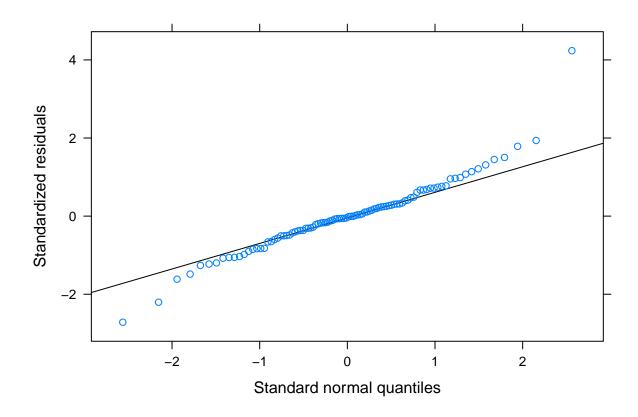
To bottom: Full analysis (all 5 groups - M chronic saline, M chronic morphine,

M acute morphine, Flo & Fhi acute morphine) - not used!!

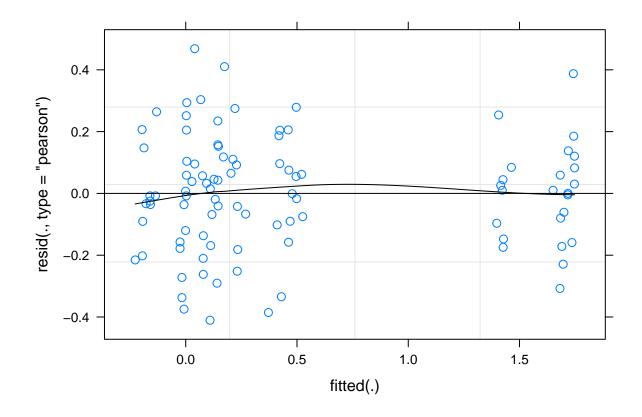
```
transporter_lm <- lmer(rge ~ gene * Tissue * Group + (1|Sample), data=transporters_long)
plot(transporter_lm, type=c("p","smooth"), col.line=1)</pre>
```



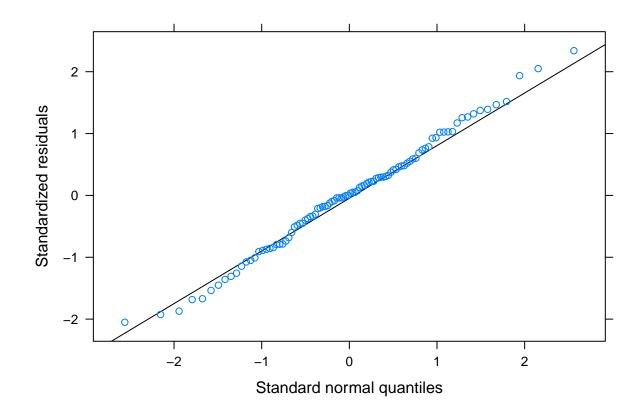
lattice::qqmath(transporter_lm)



```
 transporter_log2_lm <- lmer(log2(rge) ~ gene * Tissue * Group + (1|Sample), \\ \frac{data=}{data=} transporters_long) \\ plot(transporter_log2_lm, \\ \frac{type=}{data=} transporter_long) \\ plot(transporter_long2_lm, \\ \frac{type=}{data=} transporter_long) \\ plot(transporter_long2_lm, \\ \frac{type=}{data=} transporter_long) \\ plot(transporter_long2_lm, \\ \frac{type=}{data=} transporter_long2_lm, \\ \frac{type=}{data=}
```



lattice::qqmath(transporter_log2_lm)



```
shapiro.test(transporters_long$rge)

##
## Shapiro-Wilk normality test
##
## data: transporters_long$rge
## W = 0.77877, p-value = 1.044e-10

shapiro.test(log2(transporters_long$rge))

##
## Shapiro-Wilk normality test
##
## data: log2(transporters_long$rge)
## W = 0.87148, p-value = 1.285e-07
anova(transporter_log2_lm)
```

Sum Sq Mean Sq NumDF DenDF F value Pr(>F)

63 253.4878 < 2e-16 ***

63 509.4664 < 2e-16 ***

63 170.2831 < 2e-16 ***

4.2854 0.02749 *

1

1

2

Type III Analysis of Variance Table with Satterthwaite's method

10.1573 10.1573

20.4143 20.4143

0.3434 0.1717

6.8233 6.8233

##

gene

Tissue

gene:Tissue

Group

```
## gene:Group
                    0.1278 0.0639
                                      2
                                           63
                                                1.5945 0.21109
                    0.3872 0.1936
                                      2
                                           63 4.8315 0.01117 *
## Tissue:Group
## gene:Tissue:Group 0.1405 0.0703
                                           63 1.7536 0.18149
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
emmeans::emmeans(transporter_log2_lm, pairwise ~ Group | Tissue, by="gene")$contrasts
## Tissue = ret, gene = Pgp:
## contrast estimate
                         SE df t.ratio p.value
## M - Flo
             -0.1312 0.1180 83 -1.111 0.5098
## M - Fhi
             -0.1306 0.0962 83 -1.357 0.3680
## Flo - Fhi 0.0006 0.1117 83 0.005 1.0000
## Tissue = hyp, gene = Pgp:
## contrast estimate
                         SE df t.ratio p.value
## M - Flo -0.3108 0.1180 83 -2.633 0.0270
## M - Fhi
            -0.2798 0.0962 83 -2.909 0.0128
## Flo - Fhi 0.0310 0.1117 83 0.277 0.9585
## Tissue = ret, gene = Bcrp:
## contrast estimate
                         SE df t.ratio p.value
## M - Flo -0.0658 0.1180 83 -0.557 0.8430
## M - Fhi 0.1748 0.0962 83 1.818 0.1702
## Flo - Fhi 0.2406 0.1117 83 2.155 0.0852
##
## Tissue = hyp, gene = Bcrp:
## contrast estimate
                        SE df t.ratio p.value
           -0.1804 0.1180 83 -1.528 0.2830
## M - Flo
## M - Fhi
             -0.2530 0.0962 83 -2.630 0.0272
## Flo - Fhi -0.0726 0.1117 83 -0.651 0.7926
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
## P value adjustment: tukey method for comparing a family of 3 estimates
emmeans::emmeans(transporter_log2_lm, pairwise ~ Tissue | Group, by="gene")$contrasts
## Group = M, gene = Pgp:
## contrast estimate
                         SE df t.ratio p.value
## ret - hyp -1.423 0.1001 63 -14.219 <.0001
##
## Group = Flo, gene = Pgp:
## contrast estimate SE df t.ratio p.value
## ret - hyp -1.603 0.1266 63 -12.660 <.0001
##
## Group = Fhi, gene = Pgp:
## contrast estimate
                         SE df t.ratio p.value
## ret - hyp -1.572 0.0854 63 -18.422 <.0001
## Group = M, gene = Bcrp:
## contrast estimate
                         SE df t.ratio p.value
## ret - hyp -0.229 0.1001 63 -2.287 0.0255
```

```
##
## Group = Flo, gene = Bcrp:
## contrast estimate
                        SE df t.ratio p.value
## ret - hyp -0.344 0.1266 63 -2.714 0.0086
## Group = Fhi, gene = Bcrp:
## contrast estimate
                         SE df t.ratio p.value
## ret - hyp -0.657 0.0854 63 -7.695 <.0001
##
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
emmeans::emmeans(transporter_log2_lm, pairwise ~ gene | Group, by="Tissue")$contrasts
## Group = M, Tissue = ret:
## contrast
             estimate
                          SE df t.ratio p.value
## Pgp - Bcrp 0.0000 0.1001 63 0.000 1.0000
## Group = Flo, Tissue = ret:
## contrast estimate
                          SE df t.ratio p.value
## Pgp - Bcrp 0.0654 0.1266 63
                                  0.516 0.6073
##
## Group = Fhi, Tissue = ret:
## contrast estimate
                          SE df t.ratio p.value
## Pgp - Bcrp 0.3054 0.0854 63
                                  3.578 0.0007
##
## Group = M, Tissue = hyp:
                          SE df t.ratio p.value
## contrast estimate
## Pgp - Bcrp 1.1942 0.1001 63 11.932 <.0001
##
## Group = Flo, Tissue = hyp:
## contrast estimate
                          SE df t.ratio p.value
## Pgp - Bcrp 1.3247 0.1266 63 10.463 <.0001
##
## Group = Fhi, Tissue = hyp:
## contrast estimate
                           SE df t.ratio p.value
## Pgp - Bcrp 1.2211 0.0854 63 14.306 <.0001
##
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
trans_sumstats <- transporters_long %>% group_by(Group, Tissue, gene) %>% summarise(
 n = n()
 mean = mean(rge),
 sd = sd(rge),
 log2_mean = mean(log2(rge)),
 log2_sd = sd(log2(rge))
)
```

'summarise()' has grouped output by 'Group', 'Tissue'. You can override using

the '.groups' argument.

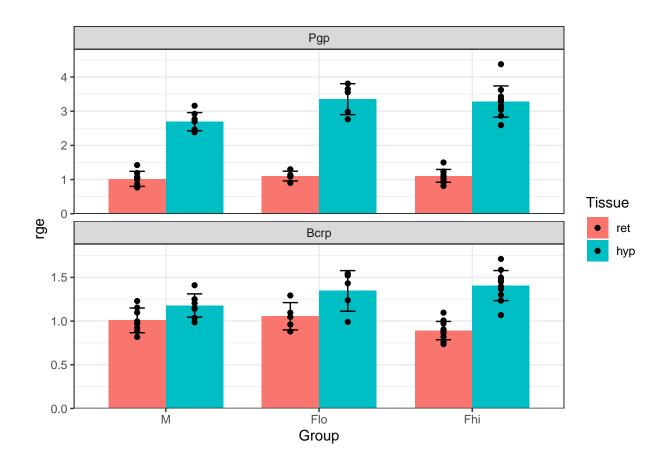
trans sumstats

```
## # A tibble: 12 x 8
## # Groups: Group, Tissue [6]
     Group Tissue gene
                          n mean
                                     sd log2_mean log2_sd
     <fct> <fct> <fct> <int> <dbl> <dbl>
##
                                            <dbl>
                                                   <dbl>
          ret
## 1 M
                           8 1.02 0.220 2.02e-10
                                                   0.305
                 Pgp
## 2 M
                           8 1.01 0.142 6.58e-11
           ret
                 Bcrp
                                                   0.202
## 3 M
           hyp
                 Pgp
                           8 2.69 0.266 1.42e+ 0
                                                   0.141
## 4 M
           hyp
                 Bcrp
                           8 1.18 0.133 2.29e- 1
                                                   0.163
## 5 Flo
           ret
                 Pgp
                           5 1.10 0.143 1.31e- 1
                                                   0.190
                           5 1.06 0.156 6.58e- 2
## 6 Flo
          ret
                 Bcrp
                                                   0.209
## 7 Flo
                           5 3.35 0.452 1.73e+ 0
                                                   0.201
          hyp
                 Pgp
## 8 Flo
           hyp
                 Bcrp
                          5 1.35 0.232 4.09e- 1
                                                   0.268
## 9 Fhi
                        11 1.11 0.187 1.31e- 1
                                                   0.242
           ret
                 Pgp
## 10 Fhi
           ret
                 Bcrp
                         11 0.891 0.105 -1.75e- 1
                                                   0.168
## 11 Fhi
                          11 3.28 0.457 1.70e+ 0
                                                   0.193
           hyp
                 Pgp
## 12 Fhi
                         11 1.41 0.172 4.82e- 1
           hyp
                 Bcrp
                                                   0.182
```

Comparing hyp vs ret within each treatment/gene

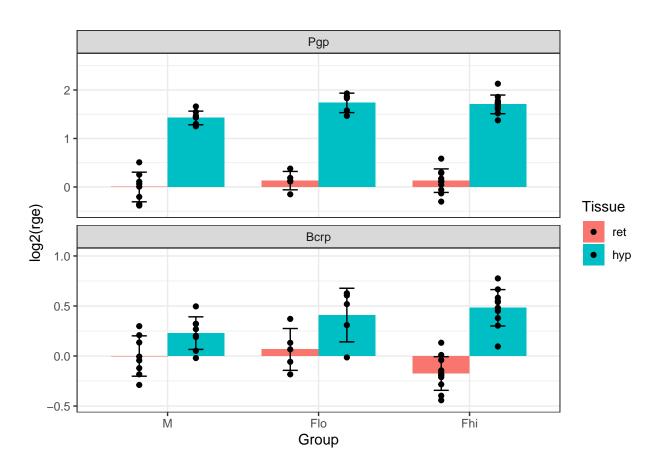
Warning: Ignoring unknown aesthetics: fill

trans_hyp_ret_plot



```
## Warning: Ignoring unknown aesthetics: fill
```

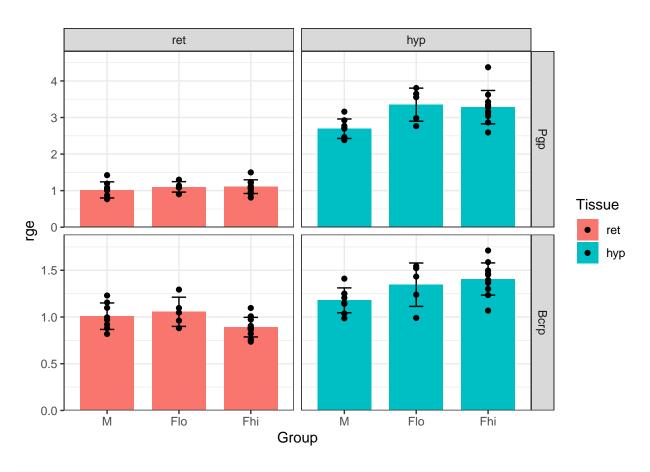
```
trans_hyp_ret_log_plot
```



 $\#ggsave(filename="../../figures/qRT-PCR/trans_hyp_ret_log.png", plot=trans_hyp_ret_log_plot, height=4,\\ \#ggsave(filename="../../figures/qRT-PCR/trans_hyp_ret_log.svg", plot=trans_hyp_ret_log_plot, height=4,\\ \#ggsave(filename="../../figures/qRT-PCR/trans_hyp_ret_log.svg", plot=trans_hyp_ret_log_plot, height=4,\\ \#ggsave(filename=".../.../figures/qRT-PCR/trans_hyp_ret_log.svg", plot=trans_hyp_ret_log_plot, height=4,\\ \#ggsave(filename=".../.../figures/qRT-PCR/trans_hyp_ret_log_plot, height=4,\\ \#ggsave(filename=".../.../figures/qRT-PCR/trans_hyp_ret_log_plot,$

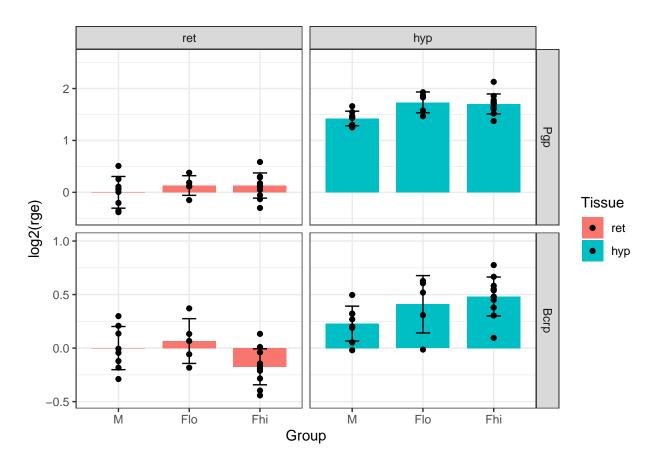
Comparing treatments within each tissue/gene

```
trans_trt_plot
```



```
## Warning: Ignoring unknown aesthetics: fill
```

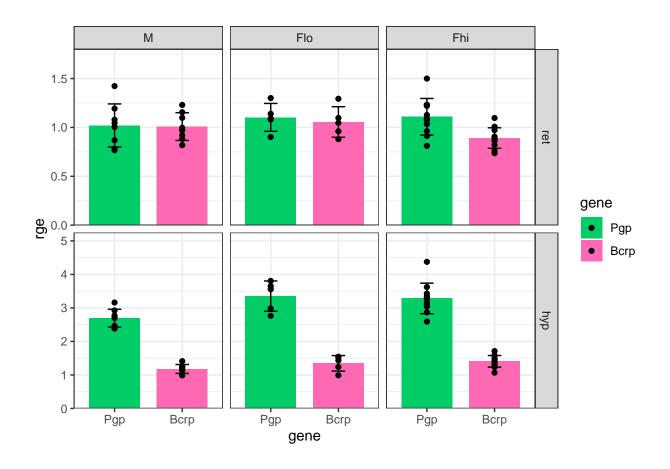
```
trans_trt_log_plot
```



 $\#ggsave(filename="../../figures/qRT-PCR/trans_trt_log.png", plot=trans_trt_log_plot, height=4, width=6) \\ \#ggsave(filename="../../figures/qRT-PCR/trans_trt_log.svg", plot=trans_trt_log_plot, height=4, width=6) \\$

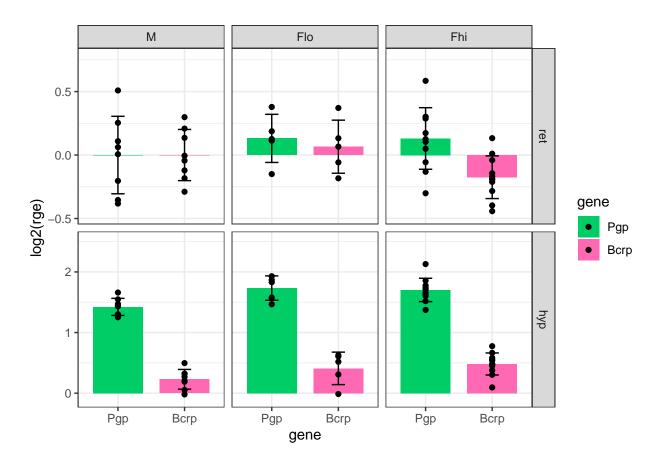
Comparing P-gp to Bcrp

```
trans_g2g_plot
```



```
## Warning: Ignoring unknown aesthetics: fill
```

```
trans_g2g_log_plot
```



 $\#ggsave(filename="../../figures/qRT-PCR/trans_g2g_log.png", plot=trans_g2g_log_plot, height=4, width=6)\\ \#ggsave(filename="../../figures/qRT-PCR/trans_g2g_log.svg", plot=trans_g2g_log_plot, height=4, width=6)\\$