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)
transporters <- read.csv(".../../data/qRT-PCR/07222022-transporter-gene-study.csv", fileEncoding = 'UTF-
transporters <- transporters %>% filter(Sample != "calib")
transporters <- transporters %% mutate(Tissue = str_sub(Sample, start=-3L, end=-1L),
                                        Treatment = str_extract(Sample, "[:alpha:]+(?=[:digit:])"),
                                        Sample = str_extract(Sample, "[:alnum:]+(?=-)"))
transporters <- transporters %>% mutate(Treatment = factor(Treatment, c("S", "M", "Flo", "Fhi", "C")),
                                        Tissue = factor(Tissue, c("ret", "hyp")))
head(transporters)
##
              Pgp.RGE Bcrp.RGE Tissue Treatment
## 1
        C1 1.7701376 1.6160126
                                   hyp
                                               С
## 2
         C1 0.4207643 0.8918919
                                   ret
                                               С
## 3
        C2 1.7514899 1.6263272
                                   hyp
        C2 0.9387229 1.2154637
                                               С
## 4
                                   ret
                                               С
## 5
         C3 1.2344328 2.3715652
                                   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)

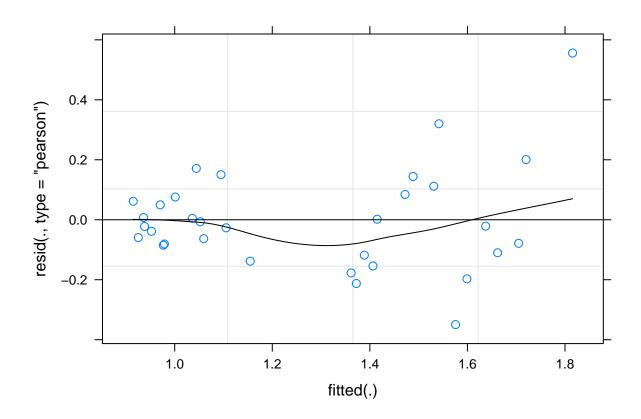
```
## # A tibble: 6 x 5
##
     Sample Tissue Treatment gene
                                       rge
##
     <chr>>
            <fct>
                   <fct>
                               <fct> <dbl>
## 1 C1
            hyp
                    С
                                     1.77
      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

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

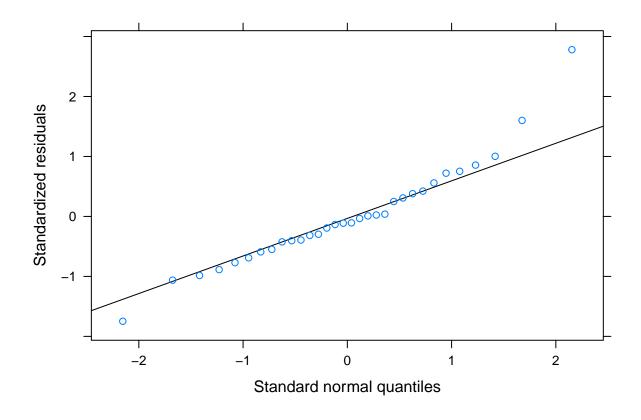
```
#bcrp only
chronic <- transporters %>% filter(Treatment %in% c("C", "S"))
chronic
```

```
##
               Pgp.RGE Bcrp.RGE Tissue Treatment
      Sample
## 1
          C1 1.7701376 1.6160126
                                                  C
                                     hyp
                                                  С
## 2
          C1 0.4207643 0.8918919
                                     ret
                                                  С
## 3
          C2 1.7514899 1.6263272
                                     hyp
## 4
                                                  С
          C2 0.9387229 1.2154637
                                     ret
## 5
          C3 1.2344328 2.3715652
                                     hyp
                                                  C
## 6
          C3 0.7343906 1.0166798
                                                  С
                                     ret
## 7
          C4 1.8650058 1.9206809
                                                  С
                                     hyp
                                                  С
## 8
          C4 0.6869529 0.9963086
                                     ret
## 9
          C5 1.3727992 1.2261280
                                                  С
                                     hyp
                                                  С
          C5 1.2258085 0.9766218
## 10
                                     ret
## 11
          C6 1.4958638 1.5518807
                                                  C
                                     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
                                                  S
## 18
          S2 0.8222355 1.0784274
                                     ret
                                                  S
## 19
          S3 1.5582238 1.1842034
                                     hyp
## 20
                                                  S
          S3 1.0081232 0.8660578
                                     ret
## 21
          S4 1.0379047 1.2523247
                                                  S
                                     hyp
                                                  S
## 22
          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
## 26
          S6 1.0835666 0.9436934
                                                  S
                                     ret
                                                  S
## 27
          S7 2.3260412 1.6427723
                                     hyp
          S7 1.4357874 1.2453351
                                                  S
## 28
                                     ret
                                                  S
## 29
          S8 2.0696005 1.6327948
                                     hyp
## 30
          S8 1.0679296 1.0456424
                                                  S
                                     ret
                                                  S
## 31
          S9 2.4339414 1.4168757
                                     hyp
## 32
          S9 1.1287493 0.8979728
                                                  S
                                     ret
```

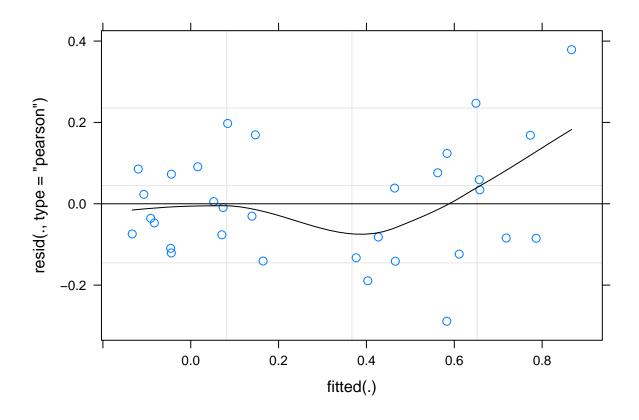
```
c_bcrp_lm <- lmer(Bcrp.RGE ~ Tissue * Treatment + (1|Sample), data=chronic)
plot(c_bcrp_lm, type=c("p","smooth"), col.line=1)</pre>
```



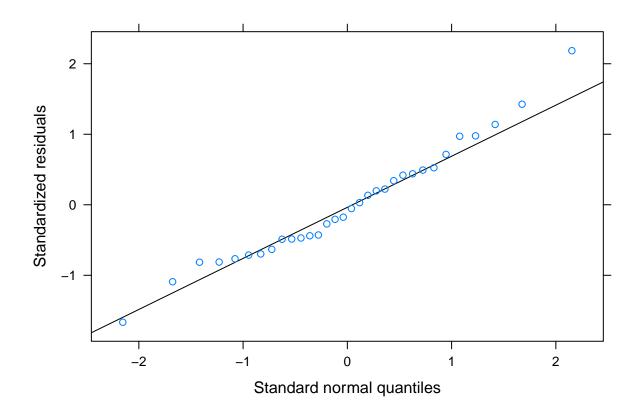
lattice::qqmath(c_bcrp_lm)



```
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>
```



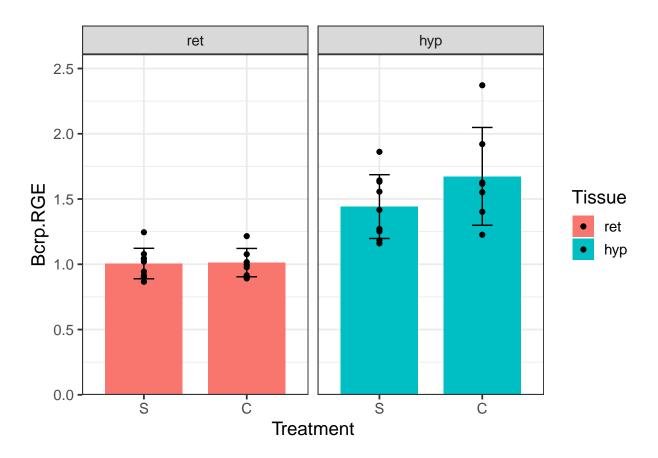
lattice::qqmath(c_bcrp_log2_lm)



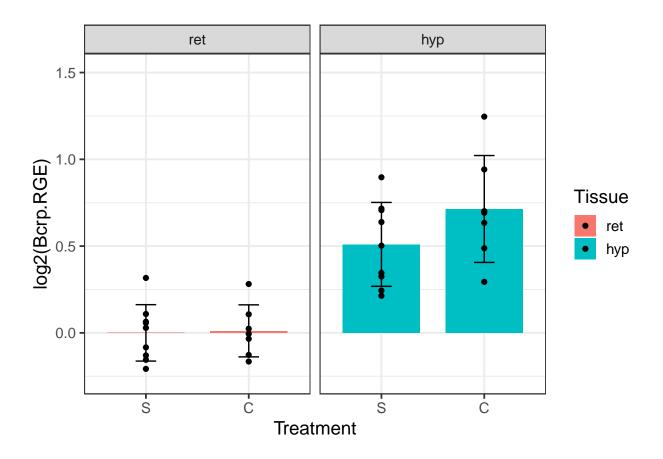
```
shapiro.test(chronic$Bcrp.RGE)
##
    Shapiro-Wilk normality test
##
##
## data: chronic$Bcrp.RGE
## W = 0.89092, p-value = 0.003667
shapiro.test(log2(chronic$Bcrp.RGE))
##
    Shapiro-Wilk normality test
##
##
## data: log2(chronic$Bcrp.RGE)
## W = 0.94026, p-value = 0.07615
anova(c_bcrp_log2_lm)
## 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)$contrasts
## Tissue = ret:
## contrast estimate
                        SE
                           df t.ratio p.value
## S - C -0.0115 0.112 24.3 -0.103 0.9188
##
## Tissue = hyp:
## contrast estimate
                           df t.ratio p.value
                        SE
## S - C
           -0.2040 0.112 24.3 -1.822 0.0808
##
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
emmeans::emmeans(c_bcrp_log2_lm, pairwise ~ Tissue | Treatment)$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),
 log2_mean = mean(log2(Bcrp.RGE)),
 log2\_sd = sd(log2(Bcrp.RGE))
)
## 'summarise()' has grouped output by 'Treatment'. You can override using the
## '.groups' argument.
c_bcrp_sumstats
## # A tibble: 4 x 7
## # Groups: Treatment [2]
    Treatment Tissue n mean
                                   sd log2_mean log2_sd
##
   <fct> <fct> <int> <dbl> <dbl>
                                          <dbl>
                                                  <dbl>
## 1 S
             ret
                        9 1.01 0.117 -4.39e-11
                                                  0.163
## 2 S
                        9 1.44 0.244 5.10e- 1
              hyp
                                                  0.242
## 3 C
                        7 1.01 0.109 1.15e- 2 0.150
             ret
## 4 C
              hyp
                       7 1.67 0.375 7.14e- 1
                                                  0.308
#treatment plots
c_bcrp_plot <- ggplot() +</pre>
```

c_bcrp_plot

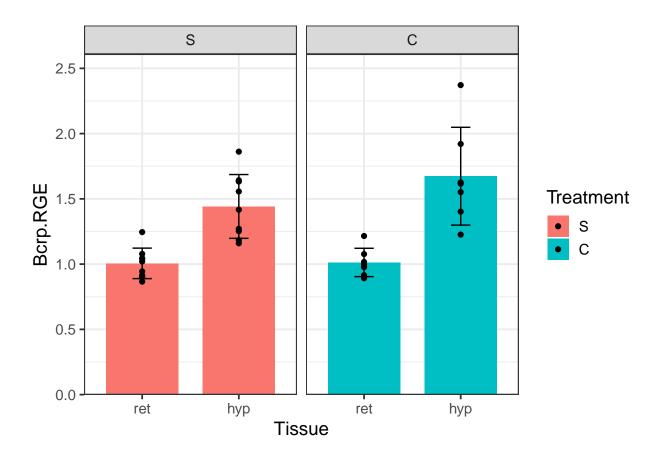


c_bcrp_log_plot

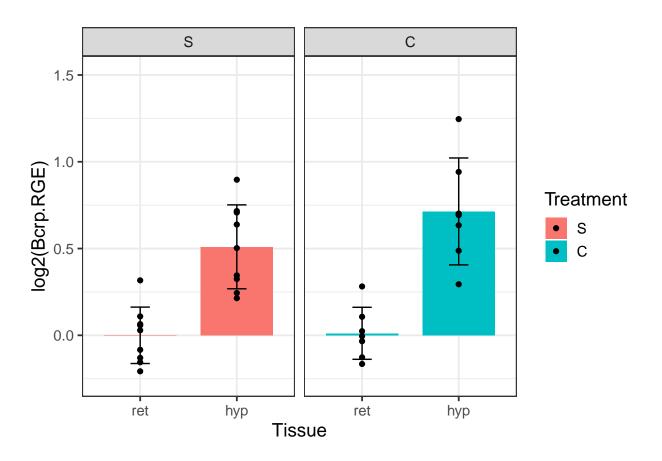


```
scale_y_continuous(expand = expansion(mult = c(0, 0.1))) +
#scale_fill_manual(values=c("maroon1", "springgreen3")) +
theme_bw(base_size = 14)
```

c_bcrp_tiss_plot



c_bcrp_tiss_log_plot

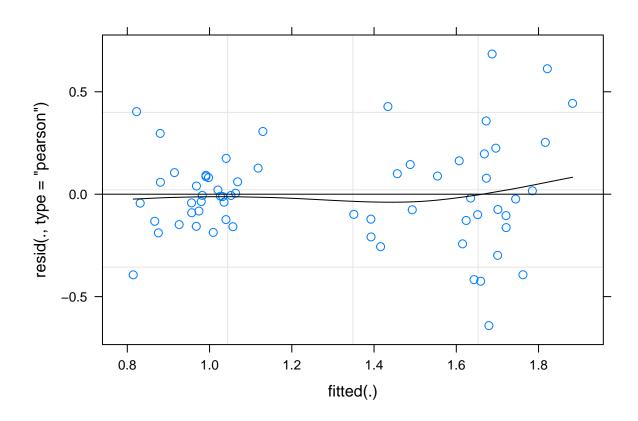


 $\#ggsave(filename=".../.../figures/qRT-PCR/c_bcrp_tiss_log_plot.svg",\ plot=c_bcrp_tiss_log_plot,\ height=4,$

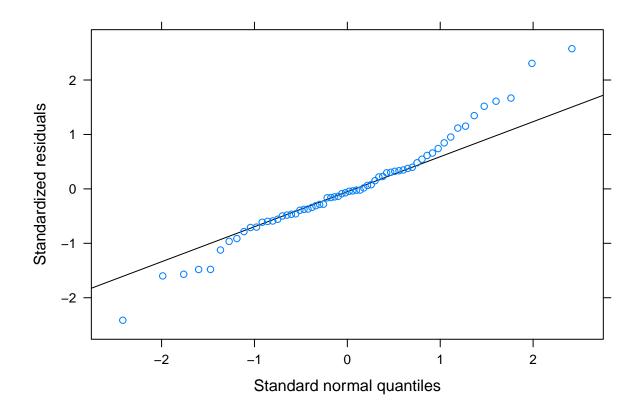
```
#bcrp & pgp together
chronic_long <- transporters_long %>% filter(Treatment %in% c("C", "S"))
chronic_long
```

```
## # A tibble: 64 x 5
##
      Sample Tissue Treatment gene
                                      rge
##
      <chr> <fct> <fct>
                              <fct> <dbl>
##
   1 C1
            hyp
                    С
                              Pgp
                                    1.77
   2 C1
                    С
                              Bcrp 1.62
##
            hyp
   3 C1
                    С
                                    0.421
##
            ret
                              Pgp
## 4 C1
                    С
                              Bcrp 0.892
            ret
## 5 C2
            hyp
                    С
                              Pgp
                                    1.75
   6 C2
                    С
                              Bcrp 1.63
##
            hyp
   7 C2
                    С
                                    0.939
##
            ret
                              Pgp
                    С
## 8 C2
            ret
                              Bcrp 1.22
## 9 C3
                    С
                                    1.23
            hyp
                              Pgp
## 10 C3
             hyp
                    С
                              Bcrp 2.37
## # ... with 54 more rows
```

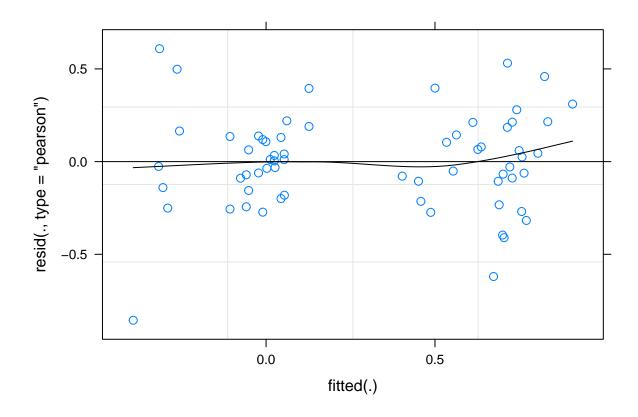
```
chronic_lm <- lmer(rge ~ gene * Tissue * Treatment + (1|Sample), data=chronic_long)
plot(chronic_lm, type=c("p","smooth"), col.line=1)</pre>
```



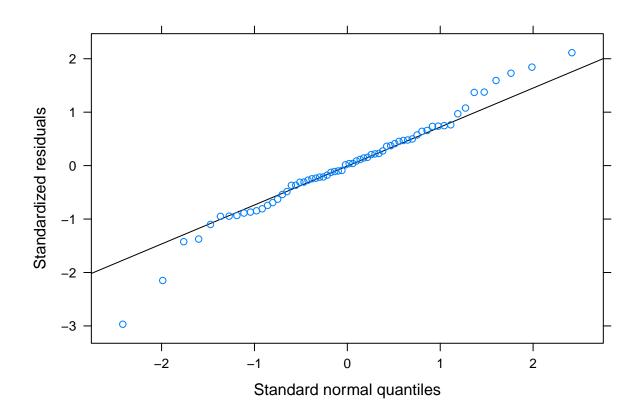
lattice::qqmath(chronic_lm)



```
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)

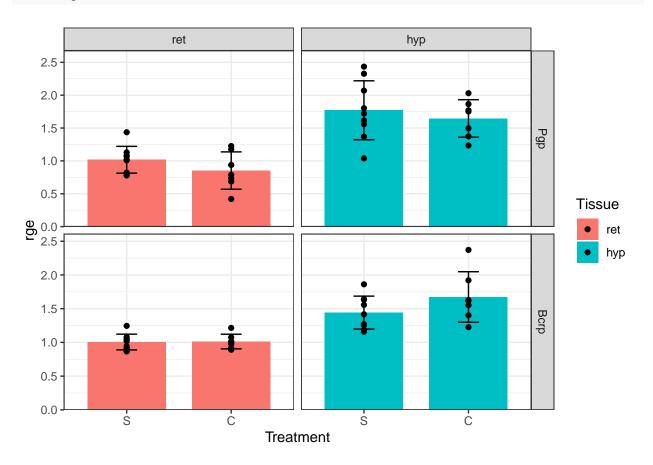
```
## Type III Analysis of Variance Table with Satterthwaite's method
##
                         Sum Sq Mean Sq NumDF DenDF F value
                                                                Pr(>F)
## gene
                         0.0038 0.0038
                                            1
                                                 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
```

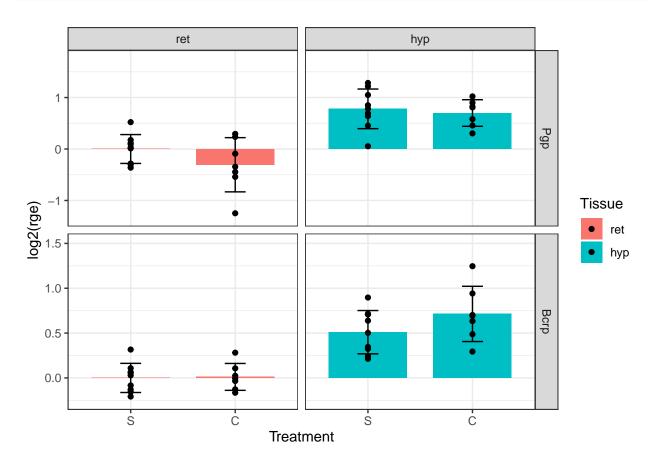
```
0.3575 0.3575 1
                                             42
                                                  4.3023 0.04423 *
## gene:Treatment
                                             42
## Tissue:Treatment
                      0.1717 0.1717
                                                  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),
 log2_mean = mean(log2(rge)),
 log2_sd = sd(log2(rge))
## 'summarise()' has grouped output by 'Treatment', 'Tissue'. You can override
## using the '.groups' argument.
chronic_sumstats
## # A tibble: 8 x 8
## # Groups: Treatment, Tissue [4]
    Treatment Tissue gene
                              n mean
                                         sd log2 mean log2 sd
##
    <fct> <fct> <fct> <int> <dbl> <dbl>
                                                <dbl>
                                                        <dbl>
## 1 S
             ret
                     Pgp
                              9 1.02 0.204 -1.11e-11
                                                        0.280
## 2 S
              ret
                     Bcrp
                               9 1.01 0.117 -4.39e-11
                                                        0.163
## 3 S
                              9 1.77 0.448 7.80e- 1
                                                        0.385
             hyp Pgp
## 4 S
             hyp Bcrp
                               9 1.44 0.244 5.10e- 1
                                                        0.242
## 5 C
                              7 0.853 0.284 -3.06e- 1
              ret Pgp
                                                        0.527
                              7 1.01 0.109 1.15e- 2
## 6 C
              ret
                     Bcrp
                                                        0.150
## 7 C
                              7 1.65 0.285 6.99e- 1
              hyp
                     Pgp
                                                        0.258
## 8 C
              hyp
                     Bcrp
                             7 1.67 0.375 7.14e- 1
                                                        0.308
chronic_plot <- ggplot() +</pre>
 stat_summary(data=chronic_long, aes(x=Treatment, y=rge, fill=Tissue),
              fun = 'mean', geom="bar", width=0.75, position="dodge") +
 geom_errorbar(data=chronic_sumstats, aes(x=Treatment, ymin=mean-sd,
                                         ymax=mean+sd, fill=Tissue),
               width=0.2, position=position_dodge(width=0.75)) +
 geom_point(data=chronic_long, aes(x=Treatment, y=rge, fill=Tissue),
            stat="identity", position=position_dodge(width=0.75)) +
 facet_grid(gene~Tissue, scales="free_y") +
```

```
scale_y_continuous(expand = expansion(mult = c(0, 0.1))) +
#scale_fill_manual(values=c("maroon1", "springgreen3")) +
theme_bw()
```

chronic_plot

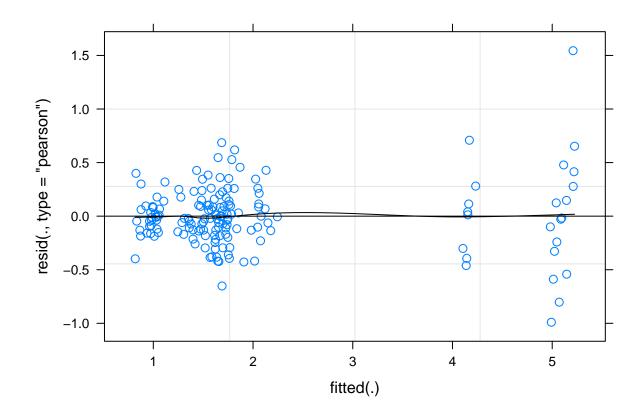




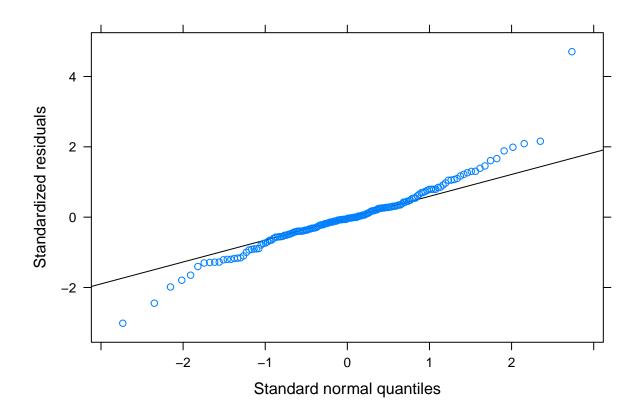
 $\#ggsave(filename=".../.../figures/qRT-PCR/chronic_trt_log.svg", ~plot=chronic_log_plot, ~height=4, ~width=6)$

Full analysis (all 5 groups) - not used

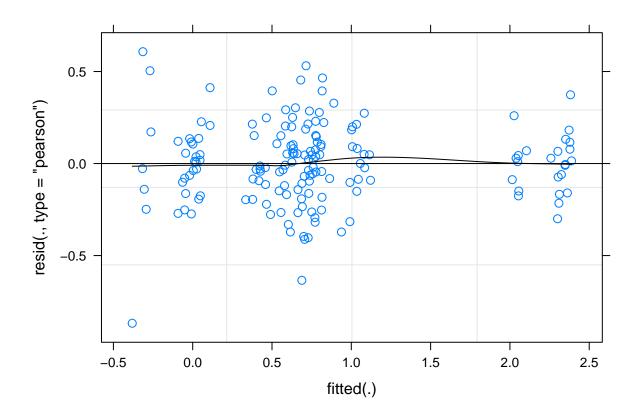
```
transporter_lm <- lmer(rge ~ gene * Tissue * Treatment + (1|Sample), data=transporters_long)
plot(transporter_lm, type=c("p","smooth"), col.line=1)</pre>
```



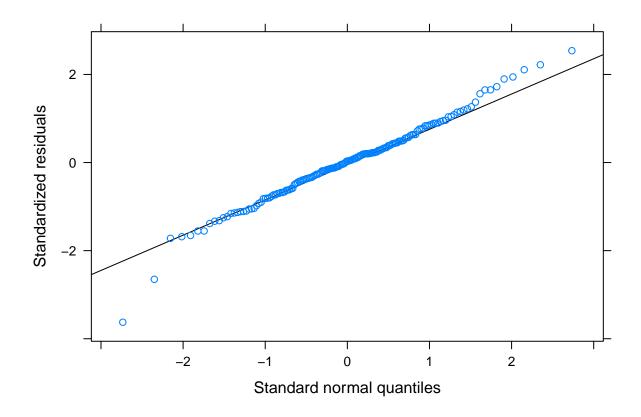
lattice::qqmath(transporter_lm)



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



lattice::qqmath(transporter_log2_lm)



```
shapiro.test(transporters_long$rge)

##
## Shapiro-Wilk normality test
##
## data: transporters_long$rge
## W = 0.75086, p-value = 3.492e-15

shapiro.test(log2(transporters_long$rge))

##
## Shapiro-Wilk normality test
##
## data: log2(transporters_long$rge)
## W = 0.92972, p-value = 4.647e-07
anova(transporter_log2_lm)
```

Sum Sq Mean Sq NumDF DenDF F value

1

Pr(>F)

105 122.0416 < 2.2e-16 ***

105 507.4983 < 2.2e-16 ***

35 77.7915 < 2.2e-16 ***
105 101.2816 < 2.2e-16 ***

Type III Analysis of Variance Table with Satterthwaite's method

6.9904 6.9904

29.0690 29.0690

17.8233 4.4558

5.8013 5.8013

##

gene

Tissue

Treatment

gene:Tissue

```
5.6966 1.4241
                                          4
                                             105
                                                  24.8634 1.699e-14 ***
## gene:Treatment
                                          4
                                             105
                                                   5.1118 0.0008386 ***
## Tissue:Treatment
                        1.1712 0.2928
## gene:Tissue:Treatment 1.6600 0.4150
                                          4
                                              105
                                                   7.2453 3.424e-05 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
emmeans::emmeans(transporter_log2_lm, pairwise ~ Treatment | Tissue, by="gene")$contrasts
## Tissue = ret, gene = Pgp:
## contrast estimate
                        SE df t.ratio p.value
## S - M
             -0.6263 0.122 136 -5.124 <.0001
## S - Flo
             -0.7575 0.140 136 -5.398 <.0001
## S - Fhi
             -0.7569 0.113 136 -6.694 <.0001
## S - C
              0.3061 0.127 136
                               2.415 0.1176
##
  M - Flo
             -0.1312 0.143 136 -0.915 0.8909
## M - Fhi
             -0.1306 0.117 136 -1.117 0.7974
              0.9324 0.130 136
                                7.162 < .0001
## M - C
## Flo - Fhi 0.0006 0.136 136
                               0.004 1.0000
## Flo - C
             1.0636 0.147 136 7.221 <.0001
## Fhi - C
              1.0630 0.122 136 8.740 <.0001
##
## Tissue = hyp, gene = Pgp:
  contrast estimate
                      SE df t.ratio p.value
             -1.2693 0.122 136 -10.383 <.0001
## S - M
   S - Flo
##
             -1.5800 0.140 136 -11.260 <.0001
##
  S - Fhi
             -1.5491 0.113 136 -13.700 <.0001
   S - C
              0.0809 0.127 136
                               0.638 0.9685
   M - Flo
##
              -0.3108 0.143 136 -2.167
                                      0.1986
## M - Fhi
             -0.2798 0.117 136 -2.394 0.1232
## M - C
              1.3502 0.130 136 10.370 <.0001
## Flo - Fhi 0.0310 0.136 136
                               0.228 0.9994
## Flo - C
              1.6610 0.147 136 11.276 <.0001
              1.6300 0.122 136 13.401 <.0001
## Fhi - C
##
## Tissue = ret, gene = Bcrp:
## contrast estimate
                        SE df t.ratio p.value
## S - M
             -0.5798 0.122 136 -4.743 0.0001
             -0.6456 0.140 136 -4.601 0.0001
## S - Flo
## S - Fhi
             -0.4050 0.113 136 -3.582 0.0043
##
   S - C
             -0.0115 0.127 136 -0.091 1.0000
## M - Flo
             -0.0658 0.143 136 -0.459 0.9908
## M - Fhi
              0.1748 0.117 136
                               1.496 0.5671
## M - C
                               4.365 0.0002
              0.5683 0.130 136
## Flo - Fhi
             0.2406 0.136 136
                               1.773 0.3934
## Flo - C
              0.6340 0.147 136
                              4.304 0.0003
##
  Fhi - C
              0.3934 0.122 136
                                3.235 0.0131
##
## Tissue = hyp, gene = Bcrp:
  contrast estimate
                        SE df t.ratio p.value
             -0.2989 0.122 136 -2.445 0.1097
## S - M
## S - Flo
             -0.4793 0.140 136 -3.416 0.0074
## S - Fhi
             -0.5519 0.113 136 -4.881 <.0001
## S - C
             -0.2040 0.127 136 -1.609 0.4941
## M - Flo -0.1804 0.143 136 -1.258 0.7174
```

```
## M - Fhi
              -0.2530 0.117 136 -2.164 0.1996
## M - C
              0.0949 0.130 136 0.729 0.9495
## Flo - Fhi -0.0726 0.136 136 -0.535 0.9835
## Flo - C
             0.2752 0.147 136 1.869 0.3393
## Fhi - C
              0.3479 0.122 136 2.860 0.0387
##
## 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 5 estimates
emmeans::emmeans(transporter_log2_lm, pairwise ~ Tissue | Treatment, by="gene")$contrasts
## Treatment = S, gene = Pgp:
## contrast estimate SE df t.ratio p.value
## ret - hyp -0.780 0.113 105 -6.916 <.0001
## Treatment = M, gene = Pgp:
## contrast estimate
                      SE df t.ratio p.value
## ret - hyp -1.423 0.120 105 -11.893 <.0001
##
## Treatment = Flo, gene = Pgp:
## contrast estimate SE df t.ratio p.value
## ret - hyp -1.603 0.151 105 -10.589 <.0001
##
## Treatment = Fhi, gene = Pgp:
## contrast estimate SE df t.ratio p.value
## ret - hyp -1.572 0.102 105 -15.408 <.0001
##
## Treatment = C, gene = Pgp:
## contrast estimate
                      SE df t.ratio p.value
## ret - hyp -1.005 0.128 105 -7.859 <.0001
## Treatment = S, gene = Bcrp:
## contrast estimate
                      SE df t.ratio p.value
## ret - hyp -0.510 0.113 105 -4.519 <.0001
## Treatment = M, gene = Bcrp:
## contrast estimate SE df t.ratio p.value
## ret - hyp -0.229 0.120 105 -1.913 0.0584
##
## Treatment = Flo, gene = Bcrp:
## contrast estimate SE df t.ratio p.value
## ret - hyp -0.344 0.151 105 -2.270 0.0253
##
## Treatment = Fhi, gene = Bcrp:
## contrast estimate
                       SE df t.ratio p.value
## ret - hyp -0.657 0.102 105 -6.436 <.0001
##
## Treatment = C, gene = Bcrp:
## contrast estimate
                      SE df t.ratio p.value
## ret - hyp -0.702 0.128 105 -5.490 <.0001
##
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
```

```
## Treatment = S, Tissue = ret:
## contrast estimate
                         SE df t.ratio p.value
## Pgp - Bcrp 0.0000 0.113 105
                                  0.000 1.0000
##
## Treatment = M, Tissue = ret:
## contrast estimate
                         SE df t.ratio p.value
## Pgp - Bcrp 0.0465 0.120 105
                                  0.389 0.6983
##
## Treatment = Flo, Tissue = ret:
                        SE df t.ratio p.value
## contrast estimate
## Pgp - Bcrp 0.1119 0.151 105
                                  0.739 0.4614
##
## Treatment = Fhi, Tissue = ret:
  contrast estimate
                       SE df t.ratio p.value
## Pgp - Bcrp 0.3519 0.102 105
                                  3.448 0.0008
## Treatment = C, Tissue = ret:
## contrast estimate
                         SE df t.ratio p.value
## Pgp - Bcrp -0.3177 0.128 105 -2.483 0.0146
##
## Treatment = S, Tissue = hyp:
## contrast estimate
                         SE df t.ratio p.value
## Pgp - Bcrp 0.2704 0.113 105
                                  2.397 0.0183
## Treatment = M, Tissue = hyp:
## contrast estimate
                         SE df t.ratio p.value
## Pgp - Bcrp 1.2408 0.120 105 10.369 <.0001
##
## Treatment = Flo, Tissue = hyp:
## contrast estimate
                        SE df t.ratio p.value
## Pgp - Bcrp 1.3712 0.151 105
                                  9.059 <.0001
##
## Treatment = Fhi, Tissue = hyp:
## contrast estimate
                         SE df t.ratio p.value
## Pgp - Bcrp 1.2676 0.102 105 12.421 <.0001
##
## Treatment = C, Tissue = hyp:
## contrast estimate
                         SE df t.ratio p.value
## Pgp - Bcrp -0.0146 0.128 105 -0.114 0.9095
##
## Degrees-of-freedom method: kenward-roger
## Results are given on the log2 (not the response) scale.
trans_sumstats <- transporters_long %>% group_by(Treatment, Tissue, gene) %>% summarise(
 n = n(),
 mean = mean(rge),
 sd = sd(rge),
 log2_mean = mean(log2(rge)),
 log2_sd = sd(log2(rge))
```

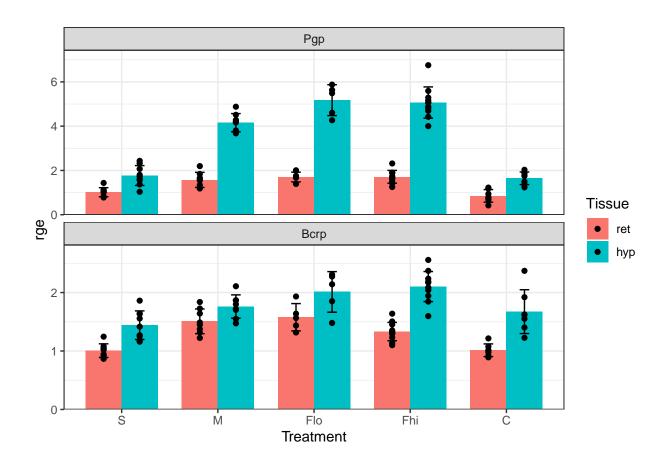
```
## 'summarise()' has grouped output by 'Treatment', 'Tissue'. You can override
## using the '.groups' argument.
```

trans_sumstats

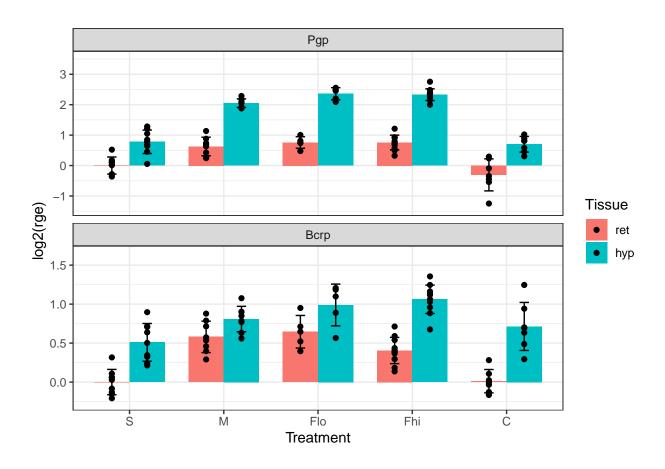
```
## # A tibble: 20 x 8
## # Groups:
              Treatment, Tissue [10]
##
     Treatment Tissue gene
                               n mean
                                          sd log2 mean log2 sd
##
               <fct> <fct> <int> <dbl> <dbl>
     <fct>
                                                 <dbl>
                                                         <dbl>
##
   1 S
                               9 1.02 0.204 -1.11e-11
                                                         0.280
               ret
                      Pgp
## 2 S
                               9 1.01 0.117 -4.39e-11
                                                        0.163
               ret
                      Bcrp
## 3 S
                     Pgp
                               9 1.77 0.448 7.80e- 1
                                                        0.385
               hyp
## 4 S
               hyp
                     Bcrp
                               9 1.44 0.244 5.10e- 1
                                                        0.242
## 5 M
                               8 1.57 0.340 6.26e- 1
                                                        0.305
               ret
                     Pgp
## 6 M
               ret
                     Bcrp
                               8 1.51 0.212 5.80e- 1
                                                        0.202
## 7 M
                               8 4.16 0.411 2.05e+ 0
                                                        0.141
               hyp
                     Pgp
## 8 M
                               8 1.76 0.199 8.09e- 1
               hyp
                     Bcrp
                                                        0.163
## 9 Flo
                               5 1.70 0.220 7.57e- 1
                                                        0.190
               ret
                     Pgp
## 10 Flo
                     Bcrp
                               5 1.58 0.234 6.46e- 1
                                                        0.209
               ret
## 11 Flo
                               5 5.17 0.698 2.36e+ 0
                                                        0.201
                     Pgp
               hyp
## 12 Flo
                               5 2.01 0.347 9.89e- 1
                                                        0.268
               hyp
                     Bcrp
## 13 Fhi
               ret
                     Pgp
                              11 1.71 0.289 7.57e- 1
                                                        0.242
## 14 Fhi
                              11 1.33 0.157 4.05e- 1
                                                         0.168
               ret
                     Bcrp
## 15 Fhi
                      Pgp
                              11 5.07 0.705 2.33e+ 0
                                                        0.193
               hyp
## 16 Fhi
                              11 2.10 0.257 1.06e+ 0
                                                        0.182
               hyp
                     Bcrp
## 17 C
                               7 0.853 0.284 -3.06e- 1
                                                        0.527
                     Pgp
               ret
## 18 C
                               7 1.01 0.109 1.15e- 2
                                                        0.150
               ret
                      Bcrp
## 19 C
                               7 1.65 0.285 6.99e- 1
                                                         0.258
               hyp
                      Pgp
## 20 C
               hyp
                      Bcrp
                               7 1.67 0.375 7.14e- 1
                                                        0.308
```

Comparing hyp vs ret within each treatment/gene

```
trans_hyp_ret_plot
```



```
trans_hyp_ret_log_plot
```

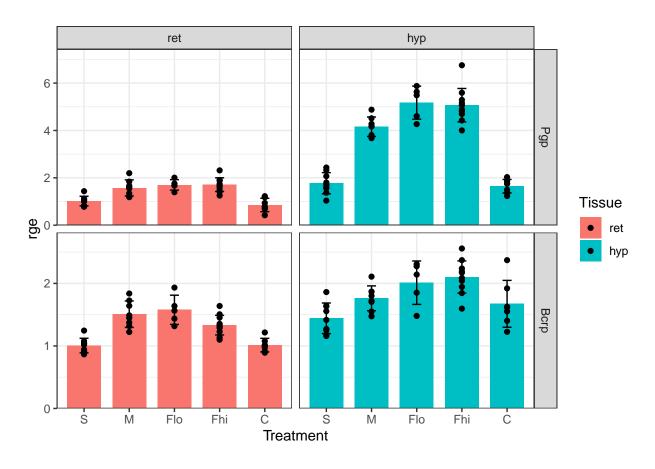


 $\#ggsave(filename="../../figures/qRT-PCR/trans_hyp_ret_log.png", plot=trans_hyp_ret_log_plot, height=4,\\ \#ggsave(filename="../../figures/qRT-PCR/trans_hyp_ret_log.svg", plot=trans_hyp_ret_log_plot, height=4,\\ \#ggsave(filename="../../figures/qRT-PCR/trans_hyp_ret_log.svg", plot=trans_hyp_ret_log_plot,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

Comparing treatments within each tissue/gene

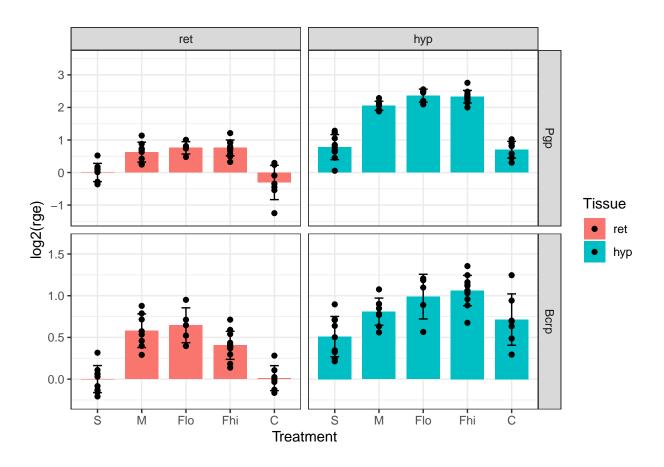
```
## Warning: Ignoring unknown aesthetics: fill
```

```
trans_trt_plot
```



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

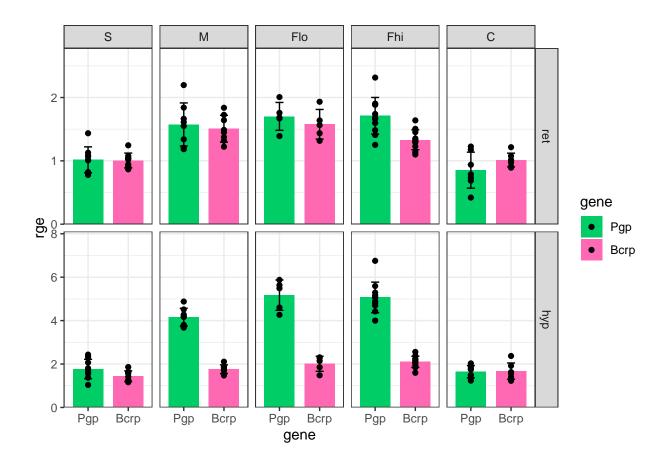
```
trans_trt_log_plot
```



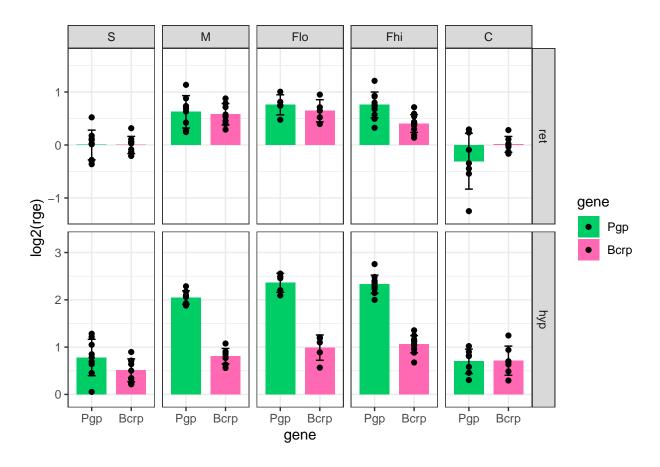
 $\#ggsave(filename="../../figures/qRT-PCR/trans_trt_log.png", plot=trans_trt_log_plot, height=4, width=6) \\ \#ggsave(filename="../../figures/qRT-PCR/trans_trt_log.svg", plot=trans_trt_log_plot, height=4, width=6) \\$

Warning: Ignoring unknown aesthetics: fill

```
trans_g2g_plot
```



```
trans_g2g_log_plot
```



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