# qpCR Transporter Figures

### C-T Berezin

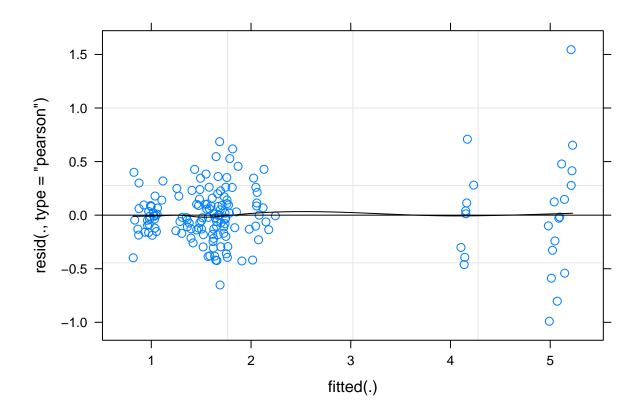
## 10/30/2021

library(tidyverse)
library(ggplot2)

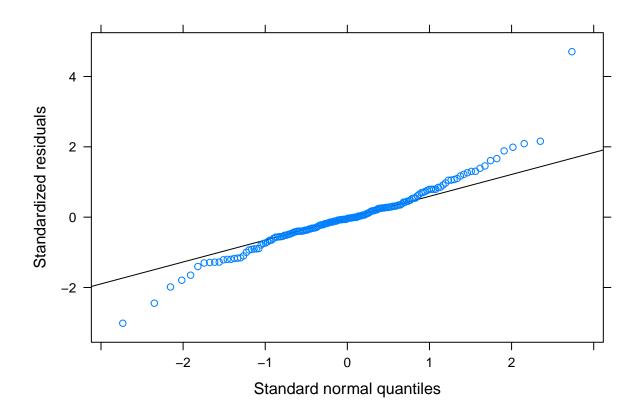
```
library(ggthemes)
library(forcats)
library(ggpubr)
#library(writexl)
library(emmeans)
library(lme4)
library(car)
library(lmerTest)
library(svglite)
transporters <- read.csv(".../data/07222022-transporter-gene-study.csv", fileEncoding = 'UTF-8-BOM')</pre>
transporters <- transporters %>% filter(Sample != "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")))
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")))
```

transporter\_lm <- lmer(rge ~ gene \* Tissue \* Treatment + (1|Sample), data=transporters\_long)</pre>

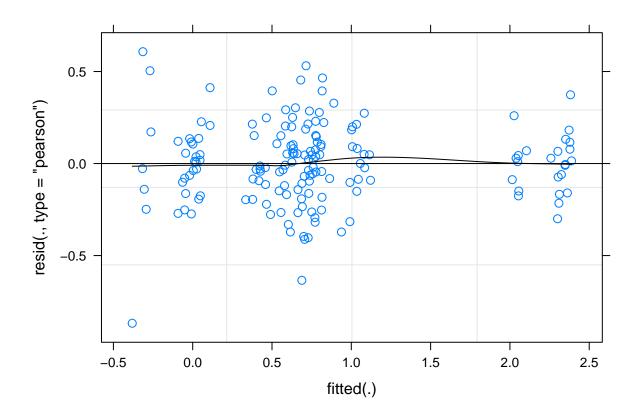
plot(transporter\_lm, type=c("p","smooth"), col.line=1)



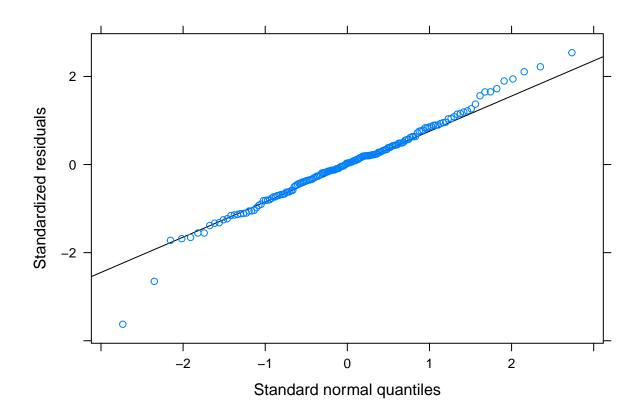
lattice::qqmath(transporter\_lm)



```
 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
## Fhi - C
              1.6300 0.122 136 13.401 <.0001
##
## 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' argu

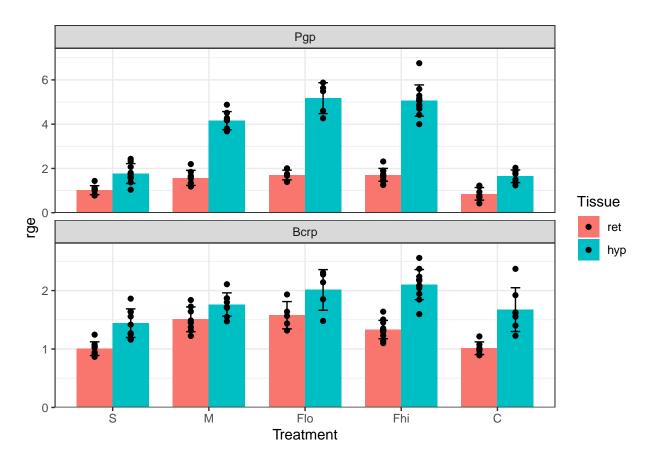
trans\_sumstats

```
## # A tibble: 20 x 8
## # Groups:
              Treatment, Tissue [10]
##
     Treatment Tissue gene
                                n mean
                                           sd log2_mean log2_sd
##
      <fct>
               <fct> <fct> <int> <dbl> <dbl>
                                                  <dbl>
                                                         <dbl>
   1 S
                                9 1.02 0.204 -1.11e-11
##
               ret
                                                         0.280
                      Pgp
                      Bcrp
                                9 1.01 0.117 -4.39e-11
   2 S
##
               ret
                                                         0.163
## 3 S
                                9 1.77 0.448 7.80e- 1
                                                         0.385
               hyp
                      Pgp
## 4 S
                                9 1.44 0.244 5.10e- 1
                                                         0.242
               hyp
                      Bcrp
## 5 M
               ret
                      Pgp
                                8 1.57 0.340 6.26e- 1
                                                         0.305
## 6 M
                                8 1.51 0.212 5.80e- 1
                                                         0.202
               ret
                      Bcrp
## 7 M
               hyp
                                8 4.16 0.411 2.05e+ 0
                                                         0.141
                      Pgp
## 8 M
                                8 1.76 0.199 8.09e- 1
                                                         0.163
               hyp
                      Bcrp
                                5 1.70 0.220 7.57e- 1
## 9 Flo
               ret
                      Pgp
                                                         0.190
## 10 Flo
               ret
                                5 1.58 0.234 6.46e- 1
                                                         0.209
                      {\tt Bcrp}
## 11 Flo
                      Pgp
                                5 5.17 0.698 2.36e+ 0
                                                         0.201
               hyp
## 12 Flo
                                5 2.01 0.347 9.89e- 1
                                                         0.268
               hyp
                      Bcrp
## 13 Fhi
                               11 1.71 0.289 7.57e- 1
                                                         0.242
               ret
                      Pgp
## 14 Fhi
               ret
                      Bcrp
                               11 1.33 0.157 4.05e- 1
                                                         0.168
## 15 Fhi
                               11 5.07 0.705 2.33e+ 0
                                                         0.193
               hyp
                      Pgp
## 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
               ret
                      Pgp
## 18 C
                                7 1.01 0.109 1.15e- 2
                                                         0.150
               ret
                      Bcrp
## 19 C
                                7 1.65 0.285 6.99e- 1
               hyp
                      Pgp
                                                         0.258
## 20 C
                                7 1.67 0.375 7.14e- 1
                                                         0.308
               hyp
                      Bcrp
```

## Comparing hyp vs ret within each treatment/gene

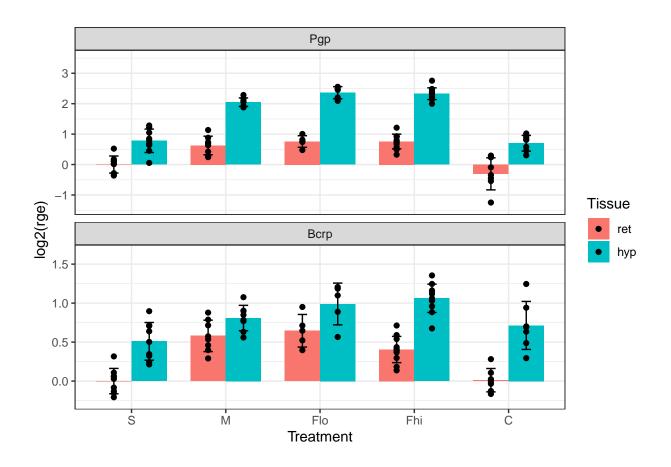
## Warning: Ignoring unknown aesthetics: fill

```
trans_hyp_ret_plot
```



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

```
trans_hyp_ret_log_plot
```

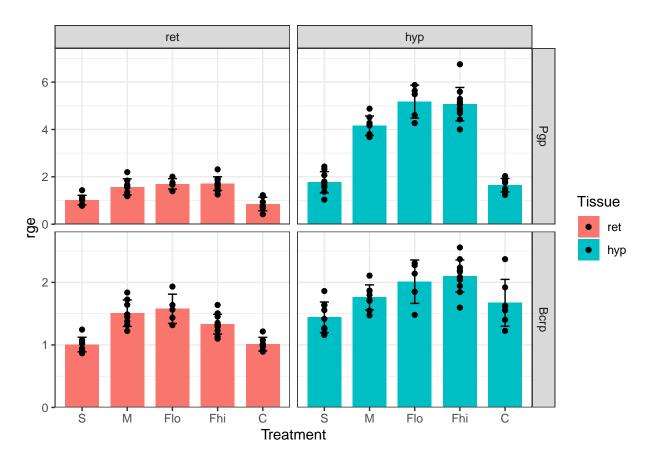


 $\#ggsave(filename="../figures/trans_hyp\_ret_log.png", plot=trans_hyp\_ret_log_plot, height=4, width=6)\\ \#ggsave(filename="../figures/trans_hyp\_ret_log.svg", plot=trans_hyp\_ret_log_plot, height=4, width=6)\\$ 

### Comparing treatments within each tissue/gene

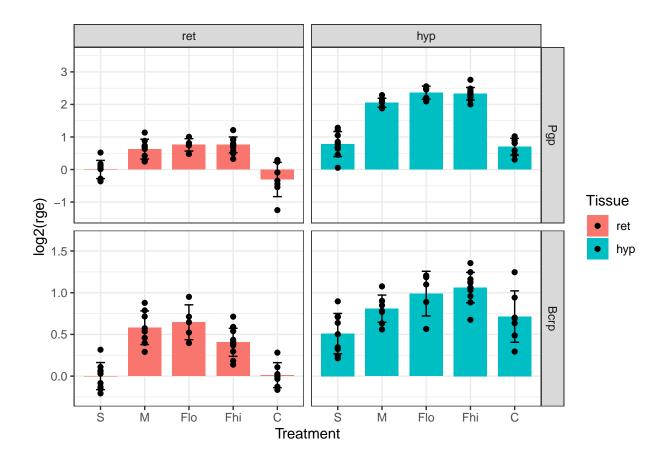
```
## Warning: Ignoring unknown aesthetics: fill
```

```
trans_trt_plot
```



## Warning: Ignoring unknown aesthetics: fill

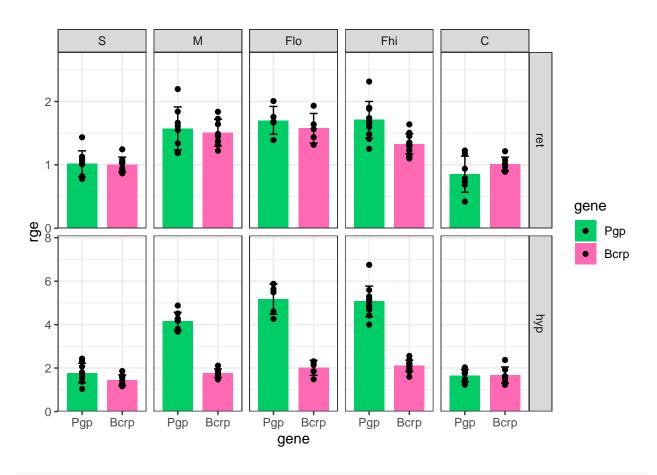
```
trans_trt_log_plot
```



 $\#ggsave(filename="../figures/trans\_trt\_log.png", plot=trans\_trt\_log\_plot, height=4, width=6)\\ \#ggsave(filename="../figures/trans\_trt\_log.svg", plot=trans\_trt\_log\_plot, height=4, width=6)\\$ 

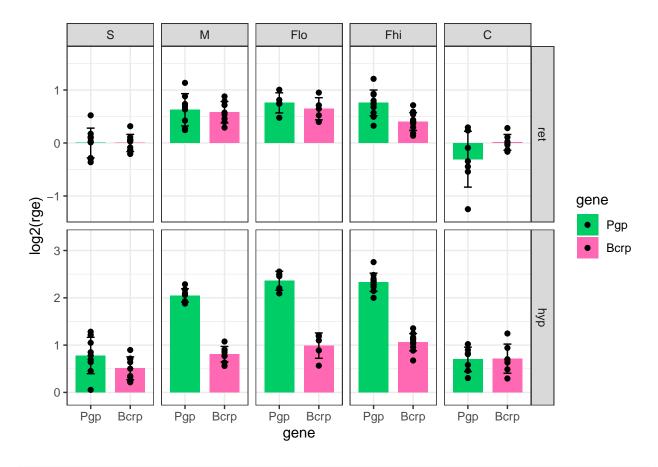
## Warning: Ignoring unknown aesthetics: fill

```
trans_g2g_plot
```



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

```
trans_g2g_log_plot
```



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