Calorimetry Plots

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R Markdown

```
rm(list=ls())
library(ggplot2)
## Warning: package 'ggplot2' was built under R version 3.5.3
library(gplots)
## Warning: package 'gplots' was built under R version 3.5.2
## Attaching package: 'gplots'
## The following object is masked from 'package:stats':
##
##
       lowess
library("dplyr")
## Warning: package 'dplyr' was built under R version 3.5.3
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
library("ggpubr")
## Loading required package: magrittr
```

```
library(tidyverse)
## -- Attaching packages ------ tidyverse 1.2.1 --
## v tibble 2.1.3 v purrr 0.3.2
## v tidyr 0.8.3 v stringr 1.4.0
## v readr 1.3.1
                    v forcats 0.4.0
## Warning: package 'tibble' was built under R version 3.5.3
## Warning: package 'tidyr' was built under R version 3.5.3
## Warning: package 'readr' was built under R version 3.5.2
## Warning: package 'purrr' was built under R version 3.5.3
## Warning: package 'stringr' was built under R version 3.5.2
## Warning: package 'forcats' was built under R version 3.5.3
## -- Conflicts ------ tidyverse_conflicts() --
## x tidyr::extract() masks magrittr::extract()
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
## x purrr::set_names() masks magrittr::set_names()
library(car)
## Warning: package 'car' was built under R version 3.5.3
## Loading required package: carData
##
## Attaching package: 'car'
## The following object is masked from 'package:purrr':
##
##
      some
## The following object is masked from 'package:dplyr':
##
##
      recode
```

```
calData <- read.csv("Calorimtery_ALX-FPR2_Calculated_Analysis_ALL_Mice_12Weeks-Old _RMoutliers_1
1.26.19.csv", header = TRUE, fill = TRUE, row.names = 1)
head(calData)</pre>
```

```
##
      Genotype Diet Groups
                                RMR Resting.Gox Resting.Lox Total.VO2
            KO Con KO_Con 299.4064
## 122
                                      8007.155
                                                 1542.3416 4434.463
## 123
            KO Con KO Con 305.4159
                                      7361.891
                                                 1189.3590 4382.593
## 191
            KO Con KO_Con 267.6917
                                      7333.698
                                                 1463.1845 3992.102
            KO Con KO_Con 264.6629
## 195
                                      6519.956
                                                 1097.3473 3621.593
## 398
            KO Con KO Con 321.5926
                                      7004.594 895.0995 4626.466
## 399
            KO Con KO Con 349.0248
                                      8036.956
                                                 1175.5389 4903.923
##
      Total.VCO2 Total.EE Total.Gox Total.Lox Total.XT VO2.Night VO2.Day
        3544.456 356.5519 8807.706 1488.834 552.9825 4755.168 3945.013
## 122
## 123
        3640.134 355.0133 8255.969 1239.907 526.0125 4660.138 3940.616
## 191
        3322.241 323.4836 7499.670 1118.668 288.2750 4115.417 3525.432
## 195
        2948.360 292.1742 7014.011 1124.298 458.6750
                                                            NaN 3245.581
## 398
        4109.049 379.7294 7860.372 864.086 741.0746 4821.000 4322.642
## 399
        4439.167 403.9906 8063.124
                                     776.143 1441.3534 5250.313 4403.623
##
      VCO2.Night VCO2.Day EE.Night EE.Day Gox.Night Gox.Day Lox.Night
        3620.203 3130.064 379.0361 316.7670 10024.392 7909.957 1897.9699
## 122
## 123
        3692.920 3222.403 374.2232 318.3346 9349.356 7585.890 1615.2545
## 191
        3329.297 2823.031 331.7581 283.5229 8038.106 6978.786 1312.8212
## 195
        2982.858 2548.445 302.5225 260.1239 7575.940 6586.885 1313.5780
## 398
        4347.135 3733.481 396.6935 352.8468 7981.248 7683.547 791.3548
        4655.970 3952.054 430.6364 362.2785 8943.263 7350.389 992.5541
## 399
        Lox.Day XT.Night
##
                           XT.Day
## 122 1363.1927 787.3375 244.3125
## 123 1199.4161 694.9188 256.8375
## 191 1173.0097 424.3687 114.7562
## 195 1164.2169 682.8375 211.1625
## 398 983.8991 918.3962 461.6667
## 399
            NaN 2186.5660 369.2830
```

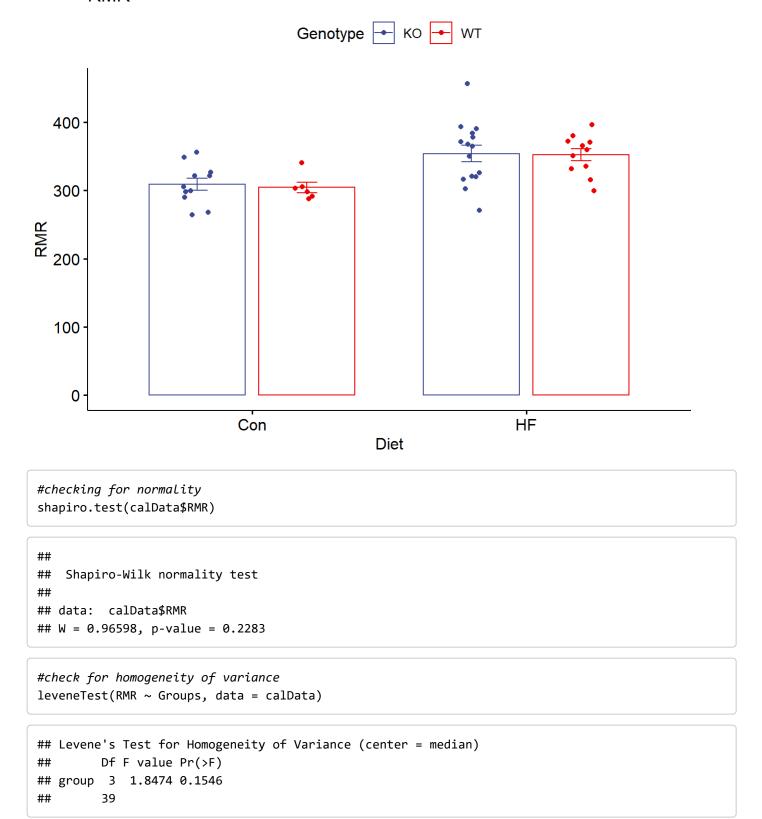
#check structure of the data to make sure that "groups" the mice belong to are factor variables str(calData)

```
'data.frame':
                   44 obs. of 24 variables:
   $ Genotype : Factor w/ 2 levels "KO", "WT": 1 1 1 1 1 1 1 1 1 1 ...
   $ Diet
                : Factor w/ 2 levels "Con", "HF": 1 1 1 1 1 1 1 1 1 1 ...
                : Factor w/ 4 levels "KO_Con", "KO_HF",..: 1 1 1 1 1 1 1 1 1 1 ...
##
   $ Groups
   $ RMR
                : num 299 305 268 265 322 ...
##
   $ Resting.Gox: num 8007 7362 7334 6520 7005 ...
##
##
   $ Resting.Lox: num 1542 1189 1463 1097 895 ...
##
   $ Total.VO2 : num 4434 4383 3992 3622 4626 ...
   $ Total.VCO2 : num 3544 3640 3322 2948 4109 ...
##
   $ Total.EE : num 357 355 323 292 380 ...
##
   $ Total.Gox : num 8808 8256 7500 7014 7860 ...
   $ Total.Lox : num 1489 1240 1119 1124 864 ...
##
   $ Total.XT : num 553 526 288 459 741 ...
   $ VO2.Night : num 4755 4660 4115 NaN 4821 ...
##
   $ VO2.Day : num 3945 3941 3525 3246 4323 ...
## $ VCO2.Night : num 3620 3693 3329 2983 4347 ...
##
   $ VCO2.Day : num 3130 3222 2823 2548 3733 ...
##
  $ EE.Night : num 379 374 332 303 397 ...
  $ EE.Day : num 317 318 284 260 353 ...
##
## $ Gox.Night : num 10024 9349 8038 7576 7981 ...
  $ Gox.Day : num 7910 7586 6979 6587 7684 ...
##
   $ Lox.Night : num 1898 1615 1313 1314 791 ...
## $ Lox.Day : num 1363 1199 1173 1164 984 ...
## $ XT.Night : num 787 695 424 683 918 ...
## $ XT.Day : num 244 257 115 211 462 ...
```

Plotting Data & Statistical Analysis

```
## Warning: Removed 1 rows containing non-finite values (stat_summary).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```



```
#Run a 2-way ANOVA
#run the multiplicative model to see if there is an interaction term (signif. diff in slope)
#my_anova <- aov(RMR ~ Diet * Genotype, data = calData)
#Anova(my_anova, type = "III") #must run a type III ANOVA because the sample size is not balance
d

#run the additive model if the interaction term is not significant
#test if there is a significant difference in the intercepts of the lines
#if there is no significant interaction effect, then type II is more powerful. If interaction is
present, then type II is inappropriate while type III can still be used.
my_anova <- aov(RMR ~ Diet + Genotype, data = calData)
Anova(my_anova, type = "III")</pre>
```

```
TukeyHSD(my_anova)
```

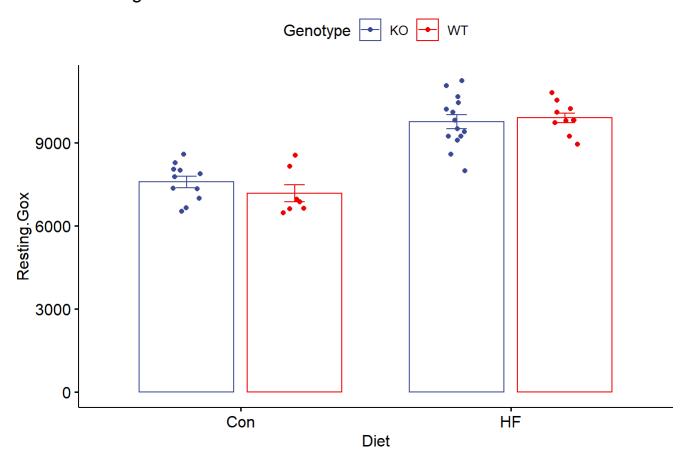
```
##
     Tukey multiple comparisons of means
##
       95% family-wise confidence level
##
## Fit: aov(formula = RMR ~ Diet + Genotype, data = calData)
##
## $Diet
##
              diff
                        lwr
                                 upr
                                         p adj
## HF-Con 45.96723 23.98914 67.94532 0.0001334
##
## $Genotype
##
              diff
                         lwr
                                upr
                                        p adj
## WT-KO -2.799084 -24.77717 19.179 0.7981896
```

The above output shows that there is only a significant difference between the dietary groups, and not the genotypes.

Warning: Removed 2 rows containing non-finite values (stat_summary).

Warning: Removed 2 rows containing missing values (geom_point).

Resting Glucose Metabolism



```
#checking for normality
shapiro.test(calData$Resting.Gox)

##
## Shapiro-Wilk normality test
##
## data: calData$Resting.Gox
## W = 0.95009, p-value = 0.06504
```

```
#check for homogeneity of variance
leveneTest(Resting.Gox ~ Groups, data = calData)
```

```
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 1.0793 0.3694
## 38
```

```
#Run a 2-way ANOVA
#run the multiplicative model to see if there is an interaction term (signif. diff in slope)
my_anova <- aov(Resting.Gox ~ Diet + Genotype, data = calData)
Anova(my_anova, type = "III") #must run a type III ANOVA because the sample size is not balanced</pre>
```

```
## Anova Table (Type III tests)
##
## Response: Resting.Gox
##
                 Sum Sq Df
                            F value
                                      Pr(>F)
## (Intercept) 789891258 1 1326.5446 < 2.2e-16 ***
## Diet
               58860004 1
                           98.8496 3.014e-12 ***
## Genotype
                  83152 1
                             0.1396
                                       0.7107
## Residuals
               23222559 39
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

```
TukeyHSD(my anova)
```

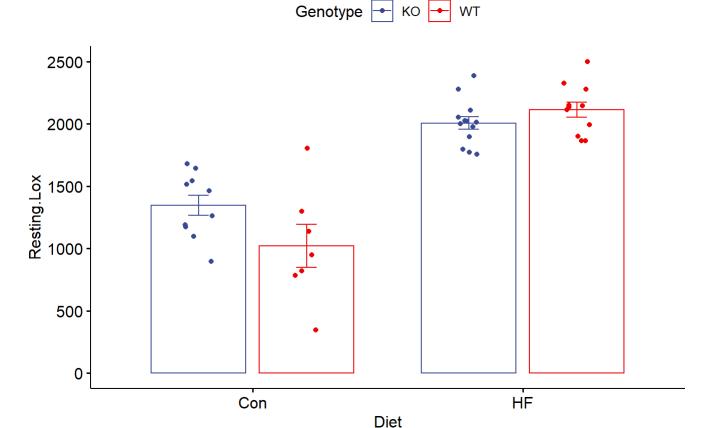
```
##
     Tukey multiple comparisons of means
##
       95% family-wise confidence level
##
## Fit: aov(formula = Resting.Gox ~ Diet + Genotype, data = calData)
##
## $Diet
##
              diff
                        lwr
                                 upr p adj
## HF-Con 2390.594 1903.924 2877.265
##
## $Genotype
##
              diff
                        lwr
                                         p adj
                                upr
## WT-KO -90.61398 -581.276 400.048 0.7107662
```

The above output shows that there is only a significant difference between the dietary groups, and not the genotypes.

```
## Warning: Removed 3 rows containing non-finite values (stat_summary).
```

```
## Warning: Removed 3 rows containing missing values (geom_point).
```

'Resting' Lipid Metabolism



```
#checking for normality
shapiro.test(calData$Resting.Lox)
```

```
##
## Shapiro-Wilk normality test
##
## data: calData$Resting.Lox
## W = 0.93766, p-value = 0.02613
```

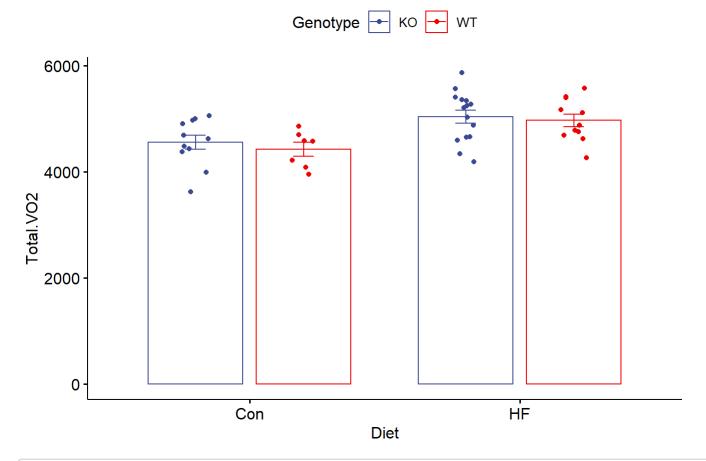
```
#check for homogeneity of variance
leveneTest(Resting.Lox ~ Groups, data = calData)
```

```
#run a wilcoxon test because data was not normal
compare_means(Resting.Lox ~ Groups, data = calData, method = "wilcox")
```

```
## # A tibble: 6 x 8
##
                 group1 group2
                                           p.adj p.format p.signif method
     .y.
##
     <chr>>
                 <chr> <chr>
                                   <dbl>
                                           <dbl> <chr>
                                                          <chr>
                                                                   <chr>>
## 1 Resting.Lox KO Con KO HF 0.0000633 0.00038 6.3e-05
                                                                   Wilcoxon
## 2 Resting.Lox KO_Con WT_Con 0.107
                                         0.21
                                                                   Wilcoxon
                                                 0.10735
                                                          ns
## 3 Resting.Lox KO Con WT HF 0.000124 0.00062 0.00012
                                                                   Wilcoxon
## 4 Resting.Lox KO_HF WT_Con 0.000874 0.0026 0.00087
                                                                   Wilcoxon
## 5 Resting.Lox KO HF WT HF 0.183
                                         0.21
                                                 0.18268
                                                                   Wilcoxon
                                                          ns
## 6 Resting.Lox WT Con WT HF 0.000578 0.0023
                                                 0.00058
                                                                   Wilcoxon
```

The data above was not normal therefore we used a wilcoxon test. No signficant differences observed between the genotypes - only a modest increase in "resting" lipid metabolism with the KO mice compared to the WTs (p = 0.1)

Total VO2



```
#checking for normality
shapiro.test(calData$Total.VO2)
```

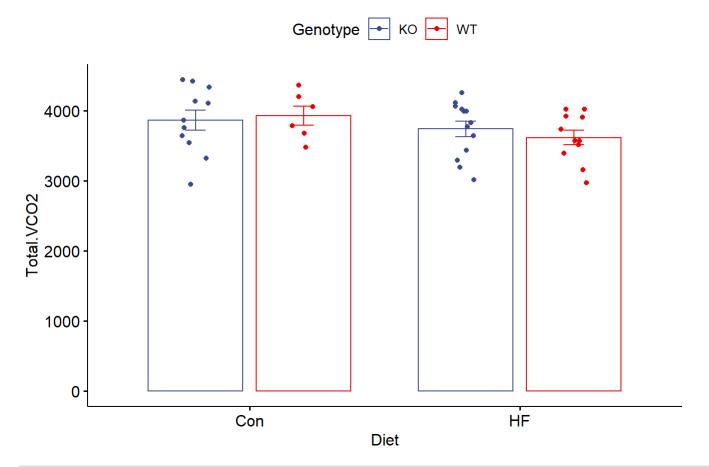
```
##
##
   Shapiro-Wilk normality test
##
## data: calData$Total.VO2
## W = 0.99179, p-value = 0.987
#check for homogeneity of variance
leveneTest(Total.VO2 ~ Groups, data = calData)
## Levene's Test for Homogeneity of Variance (center = median)
##
        Df F value Pr(>F)
## group 3 0.2698 0.8468
##
        40
#Run a 2-way ANOVA
#run the multiplicative model to see if there is an interaction term (signif. diff in slope)
my_anova <- aov(Total.VO2 ~ Diet + Genotype, data = calData)</pre>
Anova(my_anova, type = "III") #must run a type III ANOVA because the sample size is not balanced
## Anova Table (Type III tests)
##
## Response: Total.VO2
##
                 Sum Sq Df F value
                                       Pr(>F)
## (Intercept) 295983493 1 1630.6810 < 2.2e-16 ***
             2742855 1 15.1114 0.0003632 ***
## Diet
                              0.5549 0.4605743
## Genotype
                100718 1
## Residuals
                7441874 41
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
TukeyHSD(my_anova)
##
    Tukey multiple comparisons of means
      95% family-wise confidence level
##
##
## Fit: aov(formula = Total.VO2 ~ Diet + Genotype, data = calData)
##
## $Diet
             diff
##
                       lwr
                               upr
                                       p adj
## HF-Con 504.7824 240.9639 768.601 0.0003894
##
## $Genotype
             diff lwr
##
                                 upr
                                         p adj
```

WT-KO -97.25303 -361.0716 166.5655 0.4608348

Warning: Removed 3 rows containing non-finite values (stat_summary).

Warning: Removed 3 rows containing missing values (geom_point).

Total VCO2



#checking for normality
shapiro.test(calData\$Total.VCO2)

```
##
## Shapiro-Wilk normality test
##
## data: calData$Total.VCO2
## W = 0.9707, p-value = 0.3634
```

```
#check for homogeneity of variance
leveneTest(Total.VCO2 ~ Groups, data = calData)
```

```
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 0.534 0.6618
## 37
```

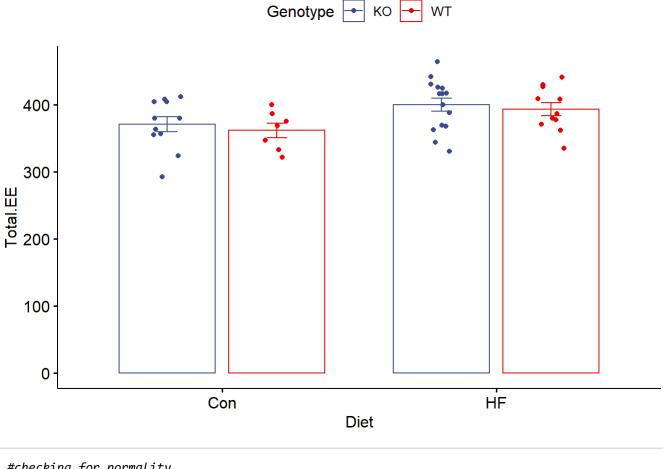
```
#Run a 2-way ANOVA
#run the multiplicative model to see if there is an interaction term (signif. diff in slope)
my_anova <- aov(Total.VCO2 ~ Diet + Genotype, data = calData)
Anova(my_anova, type = "III") #must run a type III ANOVA because the sample size is not balanced</pre>
```

```
## Anova Table (Type III tests)
##
## Response: Total.VCO2
                             F value Pr(>F)
##
                 Sum Sq Df
## (Intercept) 213300770 1 1349.4026 <2e-16 ***
## Diet
                 388699 1
                              2.4590 0.1251
## Genotype
                  25486 1
                              0.1612 0.6903
## Residuals
                6006680 38
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

TukeyHSD(my_anova)

```
##
     Tukey multiple comparisons of means
##
       95% family-wise confidence level
##
## Fit: aov(formula = Total.VCO2 ~ Diet + Genotype, data = calData)
##
## $Diet
               diff
##
                          lwr
                                   upr
                                           p adj
## HF-Con -204.1075 -459.2498 51.03471 0.113618
##
## $Genotype
##
              diff
                         lwr
                                           p adj
                                  upr
## WT-KO -50.32586 -305.4681 204.8164 0.6919044
```

Total EE



```
#checking for normality
shapiro.test(calData$Total.EE)

##
## Shapiro-Wilk normality test
##
## data: calData$Total.EE
## W = 0.9867, p-value = 0.8857

#check for homogeneity of variance
leveneTest(Total.EE ~ Groups, data = calData)
```

```
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 0.1945 0.8995
## 40
```

```
#Run a 2-way ANOVA
#run the multiplicative model to see if there is an interaction term (signif. diff in slope)
my_anova <- aov(Total.EE ~ Diet + Genotype, data = calData)
Anova(my_anova, type = "III") #must run a type III ANOVA because the sample size is not balanced</pre>
```

```
## Anova Table (Type III tests)
##
## Response: Total.EE
##
               Sum Sq Df F value
                                    Pr(>F)
## (Intercept) 1963406 1 1604.4310 < 2.2e-16 ***
## Diet
                9601 1
                           7.8460 0.007737 **
## Genotype
                 633 1
                           0.5171 0.476179
## Residuals 50173 41
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

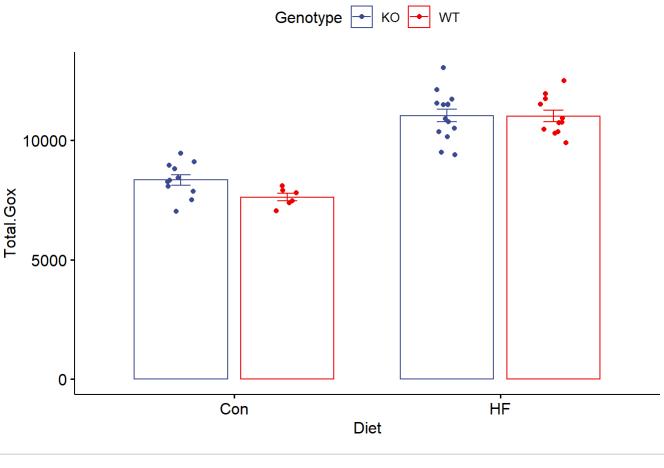
```
TukeyHSD(my_anova)
```

```
##
     Tukey multiple comparisons of means
##
       95% family-wise confidence level
##
## Fit: aov(formula = Total.EE ~ Diet + Genotype, data = calData)
##
## $Diet
              diff
##
                        lwr
                                 upr
                                         p adj
## HF-Con 29.79864 8.136539 51.46074 0.0082106
##
## $Genotype
##
             diff
                        lwr
                                 upr
                                         p adj
## WT-KO -7.70837 -29.37047 13.95373 0.4764353
```

```
## Warning: Removed 2 rows containing non-finite values (stat_summary).
```

```
## Warning: Removed 2 rows containing missing values (geom point).
```

Total Glucose Metabolism



```
#checking for normality
shapiro.test(calData$Total.Gox)

##

## Shapiro-Wilk normality test
##

## data: calData$Total.Gox

## W = 0.95275, p-value = 0.08103

#check for homogeneity of variance
leveneTest(Total.Gox ~ Groups, data = calData)
```

```
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 1.4617 0.2403
## 38
```

```
#Run a 2-way ANOVA
#run the multiplicative model to see if there is an interaction term (signif. diff in slope)
my_anova <- aov(Total.Gox ~ Diet + Genotype, data = calData)
Anova(my_anova, type = "III") #must run a type III ANOVA because the sample size is not balanced</pre>
```

```
## Anova Table (Type III tests)
##
## Response: Total.Gox
##
                 Sum Sq Df
                            F value
                                      Pr(>F)
## (Intercept) 941580275 1 1369.0716 < 2.2e-16 ***
## Diet
               88180855 1 128.2163 6.76e-14 ***
## Genotype
               866983 1
                             1.2606
                                      0.2684
## Residuals
               26822287 39
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

```
TukeyHSD(my_anova)
```

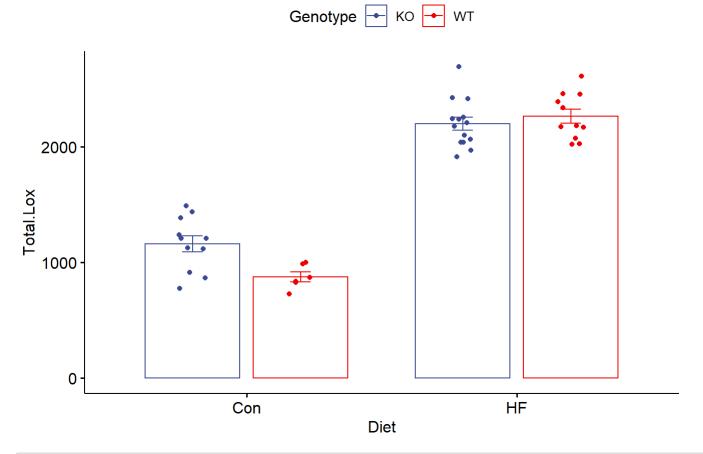
```
##
     Tukey multiple comparisons of means
##
       95% family-wise confidence level
##
## Fit: aov(formula = Total.Gox ~ Diet + Genotype, data = calData)
##
## $Diet
              diff
##
                        lwr
                                 upr p adj
## HF-Con 2937.678 2410.356 3464.999
##
## $Genotype
              diff
##
                         lwr
                                  upr
                                           p adj
## WT-KO -291.5972 -818.9185 235.7241 0.2701941
```

```
## Warning: Removed 2 rows containing non-finite values (stat_summary).
```

```
## Warning: Removed 2 rows containing missing values (geom point).
```

Total Lipid Metabolism

leveneTest(Total.Lox ~ Groups, data = calData)



```
#checking for normality
shapiro.test(calData$Total.Lox)

##

## Shapiro-Wilk normality test

##

## data: calData$Total.Lox

## W = 0.88583, p-value = 0.0005569

#check for homogeneity of variance
```

```
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 0.8271 0.4872
## 38
```

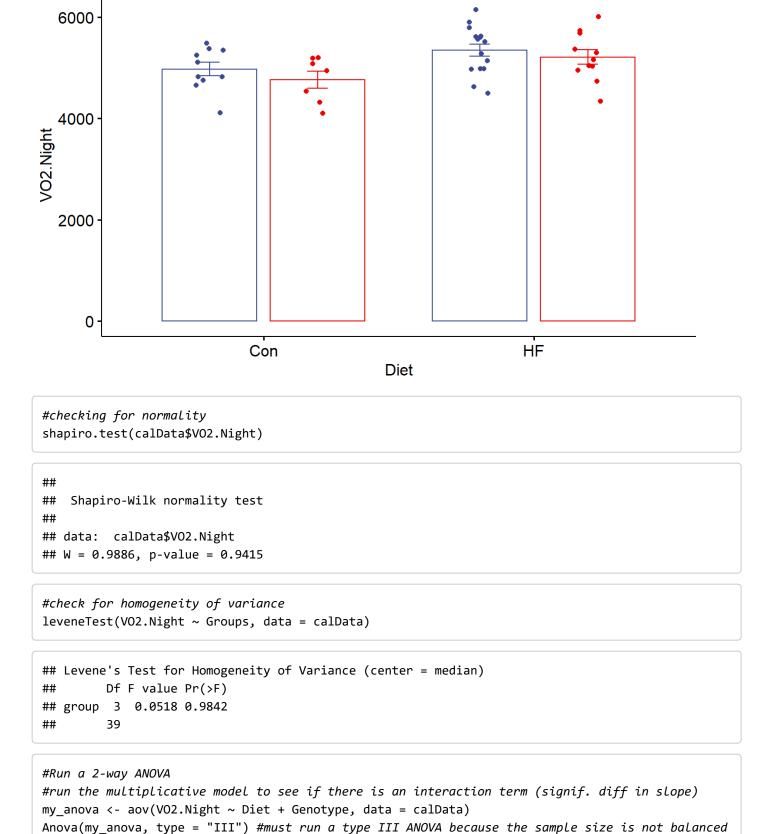
```
#run a wilcoxon test because data was not normal
compare_means(Total.Lox ~ Groups, data = calData, method = "wilcox")
```

```
## # A tibble: 6 x 8
##
                                        p.adj p.format p.signif method
     .y.
               group1 group2
                                    р
##
     <chr>>
              <chr> <chr>
                                 <dbl>
                                        <dbl> <chr>
                                                                 <chr>>
## 1 Total.Lox KO Con KO HF 0.0000281 0.00017 2.8e-05
                                                                 Wilcoxon
## 2 Total.Lox KO_Con WT_Con 0.0237
                                       0.047
                                              0.02374
                                                                Wilcoxon
## 3 Total.Lox KO Con WT HF 0.0000815 0.00041 8.2e-05
                                                                Wilcoxon
## 4 Total.Lox KO_HF WT_Con 0.000620 0.0025 0.00062
                                                                Wilcoxon
## 5 Total.Lox KO HF WT HF 0.529
                                       0.53
                                              0.52898
                                                                Wilcoxon
                                                       ns
## 6 Total.Lox WT_Con WT_HF 0.00109
                                       0.0033 0.00109
                                                                 Wilcoxon
```

The data above was not normal therefore we used a wilcoxon test. There is a signgicant difference between KO Con and WT Con. There is a significant increase in total lipid metabolism with the control KO mice compared to the WTs (p = 0.02). However this difference is lost upon a high fat diet. This indicates that at baseline the knockout of FPR2 causes the KO mice to have higher total lipid metabolism compared to the WT mice, however this effect is lost when the mice are challeged with a HF diet.

```
## Warning: Removed 1 rows containing non-finite values (stat_summary).
```

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

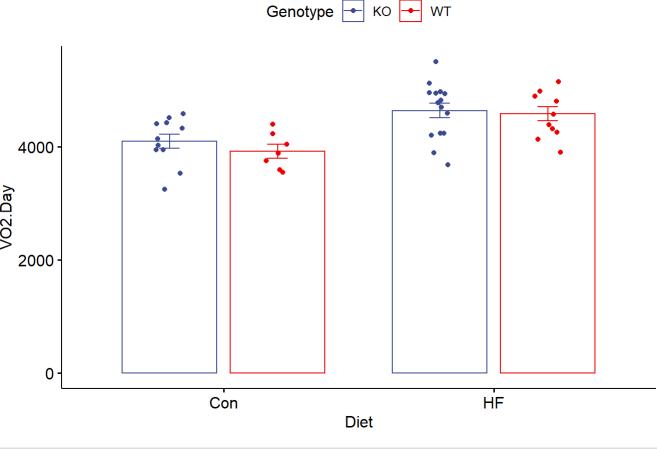


Genotype → KO →

```
## Anova Table (Type III tests)
##
## Response: VO2.Night
##
                 Sum Sq Df
                             F value
                                       Pr(>F)
## (Intercept) 327413332 1 1601.9490 < 2.2e-16 ***
## Diet
                1688056 1
                              8.2592 0.006463 **
## Genotype
                281237 1
                              1.3760 0.247718
                8175375 40
## Residuals
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
TukeyHSD(my_anova)
```

```
##
     Tukey multiple comparisons of means
       95% family-wise confidence level
##
##
## Fit: aov(formula = VO2.Night ~ Diet + Genotype, data = calData)
##
## $Diet
##
              diff
                       lwr
                                upr
                                        p adj
## HF-Con 403.4149 118.425 688.4048 0.0066856
##
## $Genotype
             diff
##
                        lwr
                                 upr
                                         p adj
## WT-KO -163.922 -446.3673 118.5233 0.2477473
```



```
#checking for normality
shapiro.test(calData$VO2.Day)

##
## Shapiro-Wilk normality test
##
## data: calData$VO2.Day
## W = 0.98439, p-value = 0.8066

#check for homogeneity of variance
leveneTest(VO2.Day ~ Groups, data = calData)
```

```
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 0.3968 0.756
## 40
```

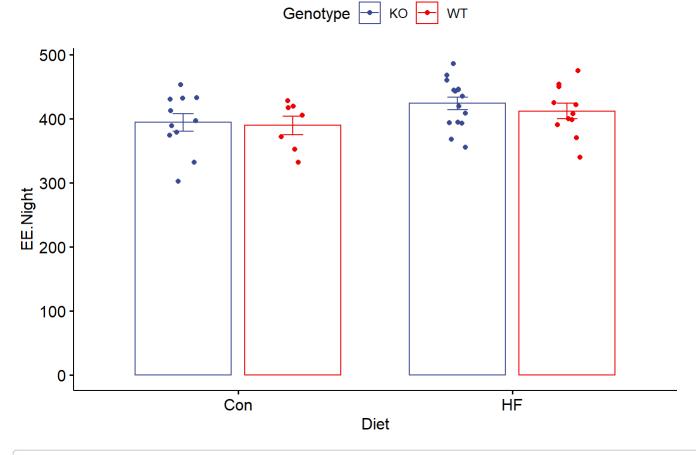
```
#Run a 2-way ANOVA
#run the multiplicative model to see if there is an interaction term (signif. diff in slope)
my_anova <- aov(VO2.Day ~ Diet + Genotype, data = calData)
Anova(my_anova, type = "III") #must run a type III ANOVA because the sample size is not balanced</pre>
```

```
## Anova Table (Type III tests)
##
## Response: VO2.Day
##
                 Sum Sq Df
                             F value
                                       Pr(>F)
## (Intercept) 237167650 1 1296.8551 < 2.2e-16 ***
## Diet
                3673754 1
                             20.0884 5.83e-05 ***
## Genotype
                119935 1
                              0.6558
                                        0.4227
## Residuals
                7498042 41
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
TukeyHSD(my_anova)
```

```
##
     Tukey multiple comparisons of means
       95% family-wise confidence level
##
##
## Fit: aov(formula = VO2.Day ~ Diet + Genotype, data = calData)
##
## $Diet
             diff
                       lwr
##
                                upr
                                       p adj
## HF-Con 584.415 319.6028 849.2273 6.31e-05
##
## $Genotype
             diff
##
                        lwr
                                 upr
                                        p adj
## WT-KO -106.126 -370.9383 158.6862 0.422986
```

EE Night Cycle



```
#checking for normality
shapiro.test(calData$EE.Night)

##
## Shapiro-Wilk normality test
##
## data: calData$EE.Night
## W = 0.98116, p-value = 0.6802

#check for homogeneity of variance
leveneTest(EE.Night ~ Groups, data = calData)
```

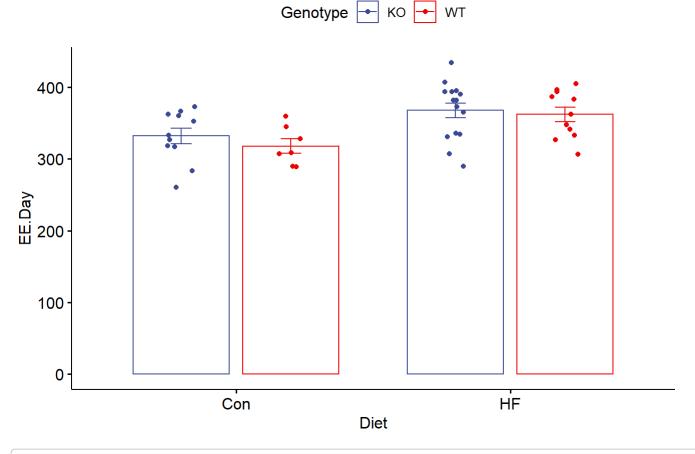
```
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 0.1061 0.9561
## 40
```

```
#Run a 2-way ANOVA
#run the multiplicative model to see if there is an interaction term (signif. diff in slope)
my_anova <- aov(EE.Night ~ Diet + Genotype, data = calData)
Anova(my_anova, type = "III") #must run a type III ANOVA because the sample size is not balanced</pre>
```

```
TukeyHSD(my_anova)
```

```
##
     Tukey multiple comparisons of means
       95% family-wise confidence level
##
##
## Fit: aov(formula = EE.Night ~ Diet + Genotype, data = calData)
##
## $Diet
##
              diff
                        lwr
                                upr
                                        p adj
## HF-Con 26.59697 2.040046 51.1539 0.0344782
##
## $Genotype
##
              diff
                         lwr
                                  upr
                                           p adj
## WT-KO -8.970644 -33.52757 15.58628 0.4648752
```

EE Day Cycle



```
#checking for normality
shapiro.test(calData$EE.Day)

##

## Shapiro-Wilk normality test

##

## data: calData$EE.Day

## W = 0.9805, p-value = 0.6536

#check for homogeneity of variance
leveneTest(EE.Day ~ Groups, data = calData)
```

```
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 0.3125 0.8162
## 40
```

```
#Run a 2-way ANOVA
#run the multiplicative model to see if there is an interaction term (signif. diff in slope)
my_anova <- aov(EE.Day ~ Diet + Genotype, data = calData)
Anova(my_anova, type = "III") #must run a type III ANOVA because the sample size is not balanced</pre>
```

```
## Anova Table (Type III tests)
##
## Response: EE.Day
##
               Sum Sq Df F value
                                    Pr(>F)
## (Intercept) 1561387 1 1261.2974 < 2.2e-16 ***
## Diet
               16102 1
                          13.0072 0.0008342 ***
                           0.7064 0.4055243
## Genotype
               874 1
## Residuals 50755 41
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

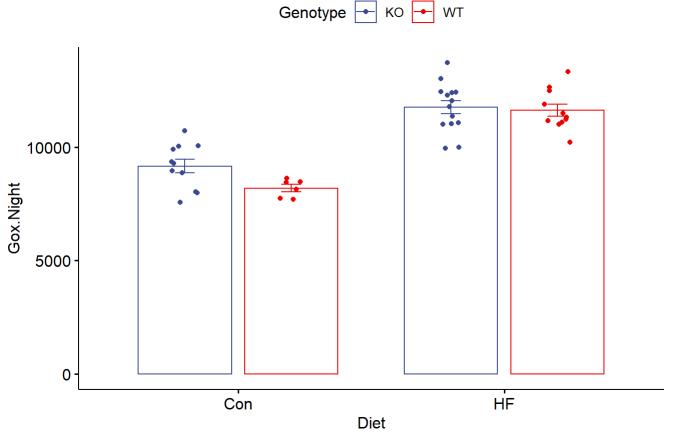
```
TukeyHSD(my_anova)
```

```
##
     Tukey multiple comparisons of means
##
       95% family-wise confidence level
##
## Fit: aov(formula = EE.Day ~ Diet + Genotype, data = calData)
##
## $Diet
              diff
##
                        lwr
                                 upr
                                          p adj
## HF-Con 38.62085 16.83359 60.40811 0.0009015
##
## $Genotype
              diff
##
                         lwr
                                  upr
                                           p adj
## WT-KO -9.061789 -30.84905 12.72547 0.4057966
```

```
## Warning: Removed 2 rows containing non-finite values (stat_summary).
```

```
## Warning: Removed 2 rows containing missing values (geom point).
```

Glucose Metabolism Night Cycle



```
#checking for normality
shapiro.test(calData$Gox.Night)

##

## Shapiro-Wilk normality test
##

## data: calData$Gox.Night

## W = 0.95728, p-value = 0.118

#check for homogeneity of variance
leveneTest(Gox.Night ~ Groups, data = calData)
```

```
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 1.3118 0.2847
## 38
```

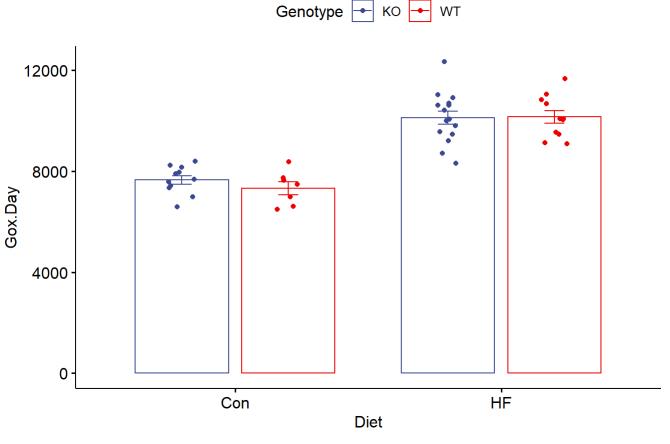
```
#Run a 2-way ANOVA
#run the multiplicative model to see if there is an interaction term (signif. diff in slope)
my_anova <- aov(Gox.Night ~ Diet + Genotype, data = calData)
Anova(my_anova, type = "III") #must run a type III ANOVA because the sample size is not balanced</pre>
```

```
## Anova Table (Type III tests)
##
## Response: Gox.Night
##
               Sum Sq Df F value
                                  Pr(>F)
## Diet
             85401002 1
                        93.9151 6.177e-12 ***
                         2.2866
## Genotype
             2079267 1
                                  0.1386
## Residuals 35464351 39
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
TukeyHSD(my_anova)
```

```
##
     Tukey multiple comparisons of means
       95% family-wise confidence level
##
##
## Fit: aov(formula = Gox.Night ~ Diet + Genotype, data = calData)
##
## $Diet
##
              diff
                        lwr
                                 upr p adj
## HF-Con 2876.562 2270.212 3482.912
##
## $Genotype
              diff
##
                         lwr
                                  upr
                                         p adj
## WT-KO -451.5785 -1057.929 154.7715 0.140023
```

Glucose Metabolism Day Cycle



```
#checking for normality
shapiro.test(calData$Gox.Day)

##
## Shapiro-Wilk normality test
##
## data: calData$Gox.Day
## W = 0.96483, p-value = 0.1967

#check for homogeneity of variance
leveneTest(Gox.Day ~ Groups, data = calData)
```

```
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 1.0903 0.3643
## 40
```

```
#Run a 2-way ANOVA
#run the multiplicative model to see if there is an interaction term (signif. diff in slope)
my_anova <- aov(Gox.Day ~ Diet + Genotype, data = calData)
Anova(my_anova, type = "III") #must run a type III ANOVA because the sample size is not balanced</pre>
```

```
## Anova Table (Type III tests)
##
## Response: Gox.Day
##
                 Sum Sq Df
                            F value
                                      Pr(>F)
## (Intercept) 822275750 1 1241.3528 < 2.2e-16 ***
## Diet
               72107468 1 108.8574 4.161e-13 ***
## Genotype
               133912 1
                             0.2022
                                       0.6554
## Residuals 27158521 41
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

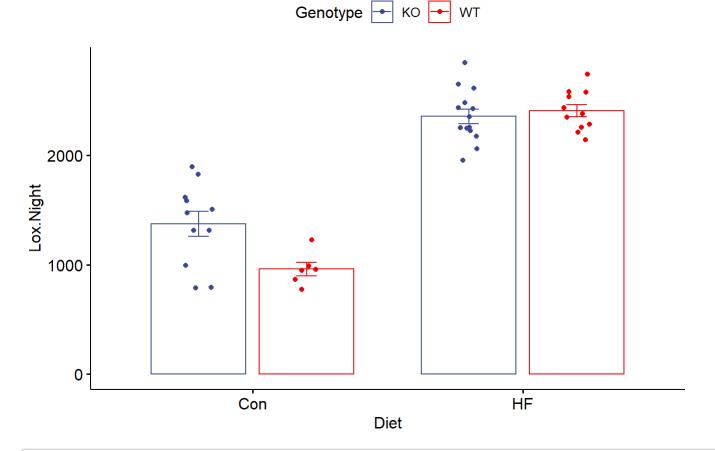
```
TukeyHSD(my_anova)
```

```
##
     Tukey multiple comparisons of means
##
       95% family-wise confidence level
##
## Fit: aov(formula = Gox.Day ~ Diet + Genotype, data = calData)
##
## $Diet
            diff
##
                      lwr
                               upr p adj
## HF-Con 2601.4 2097.416 3105.385
##
## $Genotype
              diff
##
                        lwr
                                 upr
                                         p adj
## WT-KO -112.1395 -616.124 391.8451 0.6555389
```

```
## Warning: Removed 2 rows containing non-finite values (stat_summary).
```

```
## Warning: Removed 2 rows containing missing values (geom point).
```

Lipid Metabolism Night Cycle



```
#checking for normality
shapiro.test(calData$Lox.Night)

##

## Shapiro-Wilk normality test
##

## data: calData$Lox.Night
## W = 0.89721, p-value = 0.001189
```

```
#check for homogeneity of variance
leveneTest(Lox.Night ~ Groups, data = calData)
```

```
#run a wilcoxon test because data was not normal
compare_means(Lox.Night ~ Groups, data = calData, method = "wilcox")
```

```
## # A tibble: 6 x 8
##
              group1 group2
                                        p.adj p.format p.signif method
    .y.
                                    р
##
     <chr>>
              <chr> <chr>
                                <dbl>
                                        <dbl> <chr>
                                                                <chr>>
## 1 Lox.Night KO Con KO HF 0.0000281 0.00017 2.8e-05
                                                                Wilcoxon
## 2 Lox.Night KO_Con WT_Con 0.0307
                                      0.061
                                              0.03071 *
                                                                Wilcoxon
## 3 Lox.Night KO Con WT HF 0.0000815 0.00041 8.2e-05
                                                                Wilcoxon
## 4 Lox.Night KO_HF WT_Con 0.000620 0.0025 0.00062
                                                                Wilcoxon
## 5 Lox.Night KO HF WT HF 0.565
                                      0.570
                                              0.56541
                                                                Wilcoxon
                                                      ns
## 6 Lox.Night WT_Con WT_HF 0.00109
                                      0.0033 0.00109
                                                                Wilcoxon
```

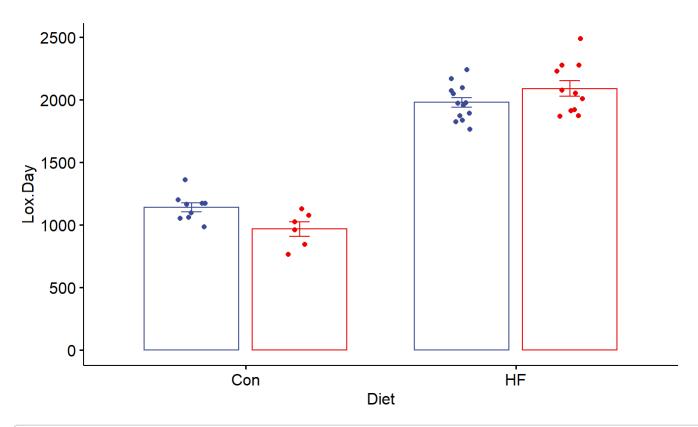
The data above was not normal therefore we used a wilcoxon test. There is a signgicant difference between KO Con and WT Con. There is a significant increase in lipid metabolism (night cycle) with the control KO mice compared to the WTs (p = 0.03). However this difference is lost upon a high fat diet. This indicates that at baseline the knockout of FPR2 causes the KO mice to have higher lipid metabolism during the night cycle (at times of activity) compared to the WT mice, however this effect is lost when the mice are challeged with a HF diet.

```
## Warning: Removed 5 rows containing non-finite values (stat_summary).
```

```
## Warning: Removed 5 rows containing missing values (geom_point).
```

Lipid Metabolism Day Cycle





```
#checking for normality
shapiro.test(calData$Lox.Day)
```

```
##
## Shapiro-Wilk normality test
##
## data: calData$Lox.Day
## W = 0.88775, p-value = 0.001002
```

```
#check for homogeneity of variance
leveneTest(Lox.Day ~ Groups, data = calData)
```

```
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 1.378 0.2656
## 35
```

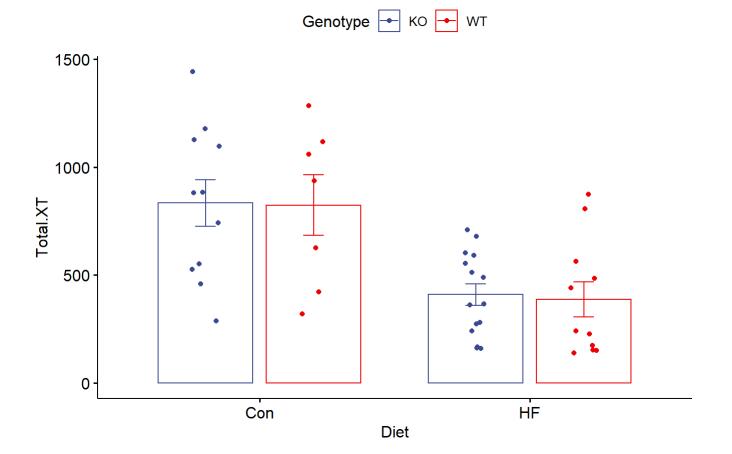
```
#run a wilcoxon test because data was not normal
compare_means(Lox.Day ~ Groups, data = calData, method = "wilcox")
```

```
## # A tibble: 6 x 8
##
             group1 group2
                                      p.adj p.format p.signif method
     .y.
                                  р
##
     <chr>>
             <chr> <chr>
                              <dbl>
                                       <dbl> <chr>
                                                      <chr>>
                                                               <chr>>
## 1 Lox.Day KO Con KO HF 0.000107 0.00064 0.00011
                                                               Wilcoxon
## 2 Lox.Day KO Con WT Con 0.0292
                                    0.058
                                                               Wilcoxon
## 3 Lox.Day KO Con WT HF 0.000197 0.00099 0.00020
                                                               Wilcoxon
## 4 Lox.Day KO HF WT Con 0.000734 0.00290 0.00073
                                                               Wilcoxon
## 5 Lox.Day KO HF WT HF
                           0.183
                                    0.18
                                             0.18268
                                                               Wilcoxon
## 6 Lox.Day WT Con WT HF 0.00109
                                    0.0033
                                            0.00109
                                                               Wilcoxon
```

The data above was not normal therefore we used a wilcoxon test. There is a signgicant difference between KO Con and WT Con. There is a significant increase in lipid metabolism (day cycle) with the control KO mice compared to the WTs (p = 0.029). However this difference is lost upon a high fat diet. This indicates that at baseline the knockout of FPR2 causes the KO mice to have higher lipid metabolism during the day cycle (at times of rest) compared to the WT mice, however this effect is lost when the mice are challeged with a HF diet. Although, there is a slight trend with the WT HF mice having elevated lipid metabolism (p = 0.18) - but it is likely not biologically meaningful.

Locomotor Analysis (XT)

Total XT



```
#checking for normality
shapiro.test(calData$Total.XT)

##
## Shapiro-Wilk normality test
##
## data: calData$Total.XT
```

```
#check for homogeneity of variance
leveneTest(Total.XT ~ Groups, data = calData)
```

```
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 1.4087 0.2544
## 40
```

```
#run a wilcoxon test because data was not normal
compare_means(Total.XT ~ Groups, data = calData, method = "wilcox")
```

```
## # A tibble: 6 x 8
##
     .у.
             group1 group2
                                 p p.adj p.format p.signif method
##
     <chr>>
             <chr> <chr>
                           <dbl> <dbl> <chr>
                                                  <chr>
                                                           <chr>>
## 1 Total.XT KO Con KO HF 0.00366 0.022 0.0037
                                                           Wilcoxon
## 2 Total.XT KO Con WT Con 1
                                   1
                                         1.0000
                                                           Wilcoxon
                                                  ns
## 3 Total.XT KO Con WT HF 0.00475 0.024 0.0047
                                                           Wilcoxon
## 4 Total.XT KO_HF WT_Con 0.0165 0.066 0.0165
                                                           Wilcoxon
## 5 Total.XT KO HF WT HF 0.406
                                   0.81 0.4063
                                                           Wilcoxon
                                                  ns
## 6 Total.XT WT_Con WT_HF 0.0185 0.066 0.0185
                                                           Wilcoxon
```

Only signficant differences between diet groups. No differences between genotypes.

W = 0.92854, p-value = 0.009293

```
Genotype → KO → WT
  2000
  1500
  1000
   500
     0
                         Con
                                                            HF
                                          Diet
#checking for normality
```

```
#checking for normality
shapiro.test(calData$XT.Night)

##

## Shapiro-Wilk normality test
##

## data: calData$XT.Night

## W = 0.91008, p-value = 0.002255

#check for homogeneity of variance
leveneTest(XT.Night ~ Groups, data = calData)
```

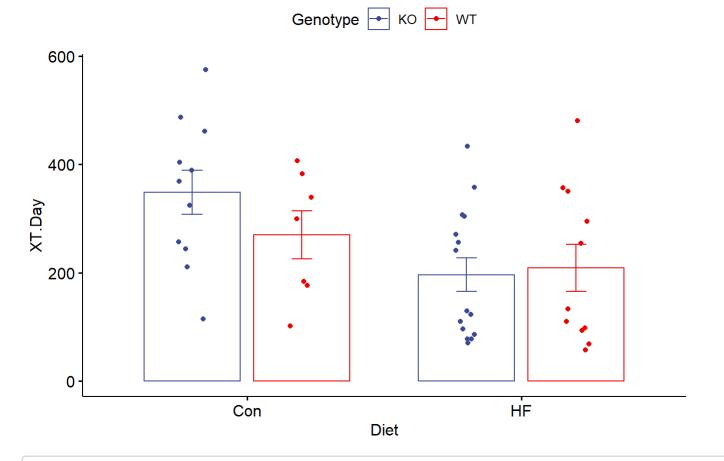
```
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 2.9888 0.04227 *
## 40
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

```
#run a wilcoxon test because data was not normal
compare_means(XT.Night ~ Groups, data = calData, method = "wilcox")
```

```
## # A tibble: 6 x 8
                                  p p.adj p.format p.signif method
##
     .у.
              group1 group2
##
     <chr>>
              <chr> <chr>
                              <dbl> <dbl> <chr>
                                                   <chr>>
                                                             <chr>>
## 1 XT.Night KO Con KO HF 0.00261 0.016 0.0026
                                                            Wilcoxon
## 2 XT.Night KO_Con WT_Con 0.717
                                    0.72 0.7172
                                                            Wilcoxon
                                                   ns
## 3 XT.Night KO_Con WT_HF 0.00386 0.019 0.0039
                                                   **
                                                            Wilcoxon
## 4 XT.Night KO_HF WT_Con 0.0112 0.045 0.0112
                                                            Wilcoxon
## 5 XT.Night KO HF WT HF 0.299
                                    0.6
                                          0.2993
                                                            Wilcoxon
                                                   ns
## 6 XT.Night WT_Con WT_HF 0.0112 0.045 0.0112
                                                            Wilcoxon
```

Only signficant differences between diet groups. No differences between genotypes.

XT Day Cycle



```
#checking for normality
shapiro.test(calData$XT.Day)
```

```
##
## Shapiro-Wilk normality test
##
## data: calData$XT.Day
## W = 0.93597, p-value = 0.01694
```

```
#check for homogeneity of variance
leveneTest(XT.Day ~ Groups, data = calData)
```

```
## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 0.1263 0.944
## 40
```

```
#run a wilcoxon test because data was not normal
compare_means(XT.Day ~ Groups, data = calData, method = "wilcox")
```

```
## # A tibble: 6 x 8
##
           group1 group2
                              p p.adj p.format p.signif method
    .у.
    <chr> <chr> <chr> <dbl> <dbl> <dbl> <chr>
                                               <chr>>
                                                       <chr>>
## 1 XT.Day KO Con KO HF 0.00946 0.057 0.0095
                                                       Wilcoxon
## 2 XT.Day KO Con WT Con 0.239
                                0.72 0.2390
                                               ns
                                                       Wilcoxon
## 3 XT.Day KO_Con WT_HF 0.0302 0.15 0.0302
                                                       Wilcoxon
## 4 XT.Day KO HF WT Con 0.159
                                0.63 0.1586
                                                       Wilcoxon
                                               ns
## 5 XT.Day KO_HF WT_HF 1
                                1
                                      1.0000
                                               ns
                                                       Wilcoxon
## 6 XT.Day WT_Con WT_HF 0.239
                                0.72 0.2390
                                               ns
                                                       Wilcoxon
```

Conclusions

Overall it seems that the FPR2 knockout alters lipid metabolism the most, throughout almost all of the lipid metabolsim calculations/timepoints. However, each time the effect is lost upon a HF diet challenge. This indicates that the gene encoding the FPR2 receptor can affect lipid metabolism under normal dietary conditions; however, in the presence of a HF diet (60% kcal) the dietary challenge of diet-induced obesity masks any effects caused by the knockout.