

Villin-Cre vs Pck1ff TG Tolerance Test

Lehninger

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Here is some old, published data from my graduate school, Am J Physiol Gastrointest Liver Physiol. 2018 Apr 6 315(2) G249–G258.

The original analysis was in Excel and GraphPad Prism, circa 2014.

The experiment tested if the knockout mouse had impaired triglyceride absorption.

```
#remember to remove the "," from the cells!
#read the CSV and load a tibble
#tg_tolerance <- read_csv("G:/My Drive/CODING - - /DataSets/AP41_wide.csv")
tg_tolerance <- read_csv("C:/Users/pottsau/Downloads/AP41_wide.csv")

## Rows: 52 Columns: 5
## — Column specification

```

```
## Delimiter: ","
## chr (1): genotype
## dbl (4): animal_num, 0hr_tg, 2hr_tg, 4hr_tg
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this
message.

head(tg_tolerance)

## # A tibble: 6 × 5
##   animal_num `0hr_tg` `2hr_tg` `4hr_tg` genotype
##   <dbl>     <dbl>   <dbl>   <dbl>   <chr>
## 1         1         73    1199    2124 Pck1ff
## 2         2         63     742    6577 Pck1ff
## 3         3        NA        NA        NA Pck1ff
## 4         4         45     955    3436 Pck1ff
## 5         5         33    1077    5876 Pck1ff
## 6         6         39    1062    3710 Pck1ff

# make the wide table into a narrow table

tg_tolerance_wide <- tg_tolerance %>%
  pivot_longer(cols = c(`0hr_tg`, `2hr_tg`, `4hr_tg`), # Specify the columns
to gather into 1 column
names_to = "time", # New column name for the old column names
```

```

        values_to = "TG_conc",
        values_drop_na = TRUE) # New column name for the values, drop
NA values

```

```
head(tg_tolerance_wide)
```

```

## # A tibble: 6 × 4
##   animal_num genotype time    TG_conc
##   <dbl> <chr>    <chr>    <dbl>
## 1         1 Pck1ff  0hr_tg     73
## 2         1 Pck1ff  2hr_tg    1199
## 3         1 Pck1ff  4hr_tg    2124
## 4         2 Pck1ff  0hr_tg     63
## 5         2 Pck1ff  2hr_tg    742
## 6         2 Pck1ff  4hr_tg   6577

```

#remember to deal with NA and NAN values! na.rm = TRUE

```

summary_stats <- tg_tolerance_wide %>%
  group_by(genotype, time) %>%
  summarize(
    mean_value = mean(TG_conc, na.rm = TRUE),
    median_value = median(TG_conc, na.rm = TRUE),
    sd_value = sd(TG_conc, na.rm = TRUE),
    min_value = min(TG_conc, na.rm = TRUE),
    max_value = max(TG_conc, na.rm = TRUE)
  )

```

`summarise()` has grouped output by 'genotype'. You can override using the
`.groups` argument.

```
print(summary_stats)
```

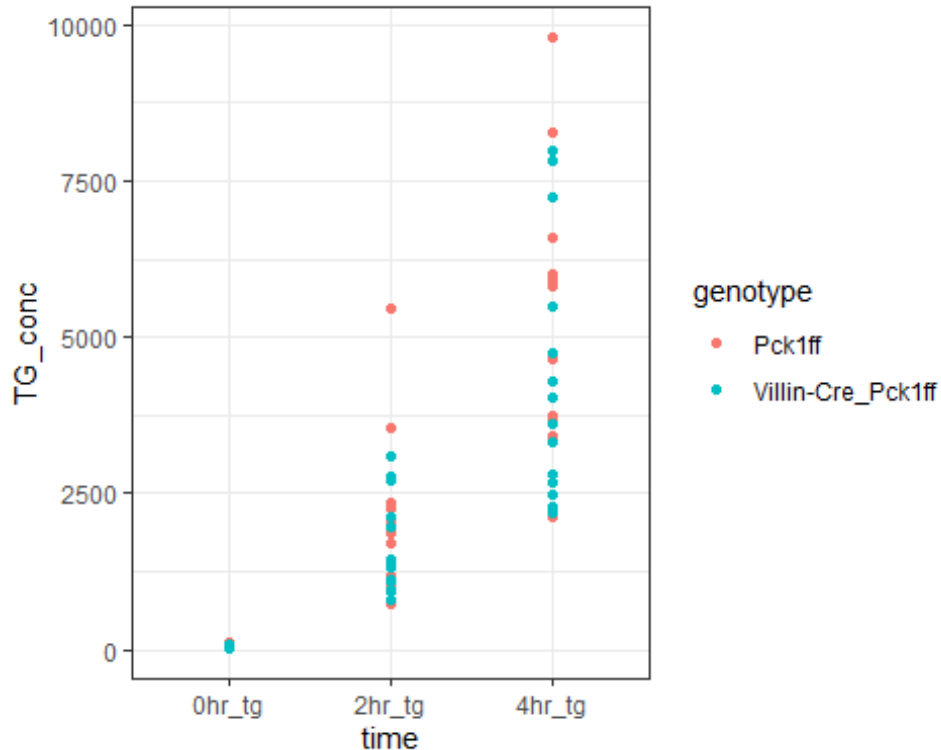
```

## # A tibble: 6 × 7
## # Groups:   genotype [2]
##   genotype      time mean_value median_value sd_value min_value
##   <chr>         <chr>    <dbl>         <dbl>    <dbl>    <dbl>
## 1 Pck1ff       0hr_tg     61.5         51.5     26.9     33
## 2 Pck1ff       2hr_tg    1606.         1077    1091.     742
## 3 Pck1ff       4hr_tg    4803.         3727    1935.    2124
## 4 Villin-Cre_Pck1ff 0hr_tg     47.2         40      15.8     29
## 5 Villin-Cre_Pck1ff 2hr_tg    1434.         1138     661.     803
## 6 Villin-Cre_Pck1ff 4hr_tg    3923.         3314    1940.    2185

```

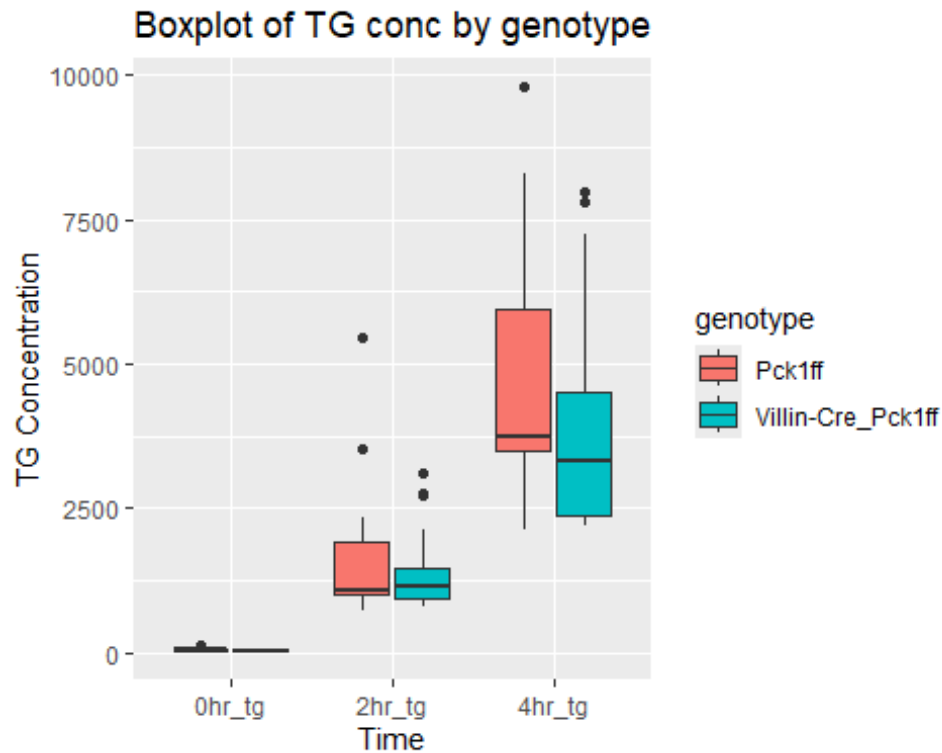
```
ggplot(data = tg_tolerance_wide) + #set data as the tg_tolerance_wide tibble
  geom_point(mapping = aes(x = time, y = TG_conc, color = genotype)) + # map
  x to time and y to TG_conc and set colorization by genotype
  geom_smooth(mapping = aes(x = time, y = TG_conc, linetype = genotype), se =
  TRUE) + # map x to time and y to TG_conc and set linetype to genotype and
  KEEP error shading w/ se
  theme_bw()

## `geom_smooth()` using method = 'loess' and formula = 'y ~ x'
```



```
## `geom_smooth()` using method = 'loess' and formula = 'y ~ x'

ggplot(tg_tolerance_wide, aes(x = time, y = TG_conc, fill = genotype)) +
  geom_boxplot() +
  labs(x = "Time", y = "TG Concentration") +
  ggtitle("Boxplot of TG conc by genotype")
```



```
# NA values removed during pivot_longer() above
```

```
# Perform two-way ANOVA
```

```
tg_ANOVA <- aov(TG_conc ~ time * genotype, data = tg_tolerance_wide)
```

```
summary(tg_ANOVA)
```

```
##              Df    Sum Sq  Mean Sq F value Pr(>F)
## time          2 429986265 214993133 141.419 <2e-16 ***
## genotype      1  4269644    4269644   2.809 0.0962 .
## time:genotype  2  4785259    2392629   1.574 0.2112
## Residuals    129 196112629    1520253
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#NORMALITY 1/3: assign residuals
```

```
#Shapiro-Wilk test: This formal test assesses the normality of the residuals.
```

```
# Extract residuals into a new column of the df
```

```
tg_tolerance_wide$residuals <- residuals(tg_ANOVA)
```

```
head(tg_tolerance_wide)
```

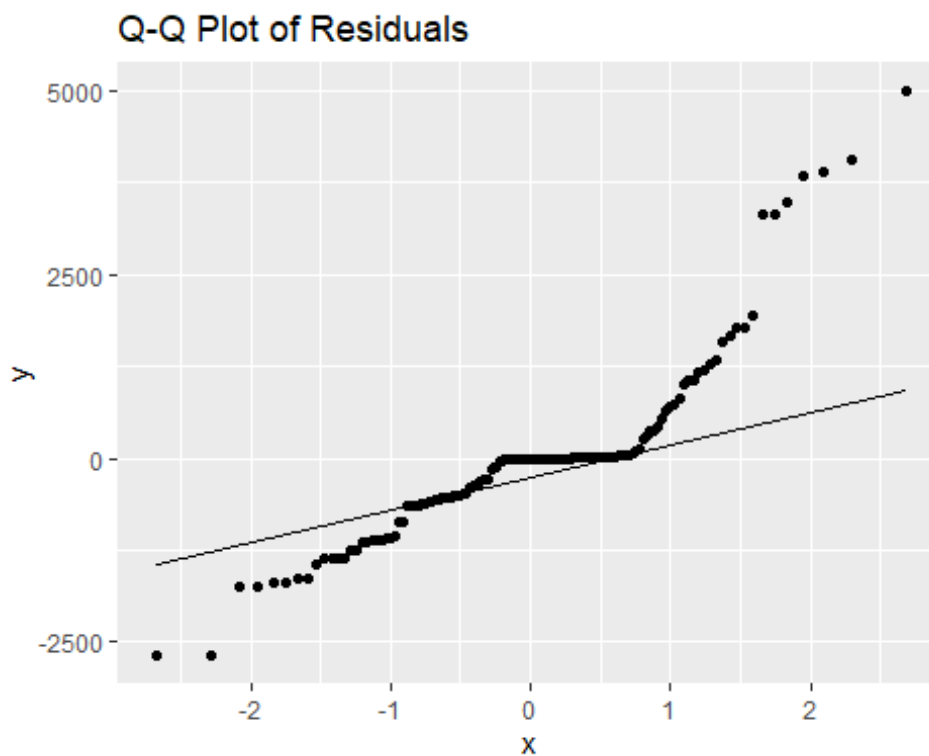
```
## # A tibble: 6 × 5
##   animal_num genotype time    TG_conc residuals
##       <dbl> <chr>   <chr>      <dbl>      <dbl>
```

```
## 1      1 Pck1ff 0hr_tg      73      11.5
## 2      1 Pck1ff 2hr_tg     1199     -407.
## 3      1 Pck1ff 4hr_tg     2124    -2679.
## 4      2 Pck1ff 0hr_tg      63       1.45
## 5      2 Pck1ff 2hr_tg      742     -864.
## 6      2 Pck1ff 4hr_tg     6577     1774.
```

#NORMALITY 2/3: Visualize Residuals

Visual inspection with QQ-plot

```
ggplot(tg_tolerance_wide, aes(sample = residuals)) +
  stat_qq() +
  stat_qq_line() +
  ggtitle("Q-Q Plot of Residuals")
```



#NORMALITY 3/3: Shapiro Wilk test

Shapiro-Wilk test for normality

```
shapiro.test(tg_tolerance_wide$residuals)
```

```
##
```

```
##  Shapiro-Wilk normality test
```

```
##
```

```
## data:  tg_tolerance_wide$residuals
```

```
## W = 0.84248, p-value = 1.047e-10
```

High p-value (e.g., $p > 0.05$): Data is likely normally distributed.

Low p-value (e.g., $p \leq 0.05$): Data is likely NOT normally distributed.

#RESULT: NOT normally distributed, try log transformation

```

# Log transform the 'TG_conc' column
tg_tolerance_wide$TG_conc_log <- log(tg_tolerance_wide$TG_conc)

# Print the transformed data
print(tg_tolerance_wide)

## # A tibble: 135 × 6
##   animal_num genotype time   TG_conc residuals TG_conc_log
##   <dbl> <chr>   <chr>   <dbl>     <dbl>     <dbl>
## 1         1 Pck1ff 0hr_tg     73      11.5      4.29
## 2         1 Pck1ff 2hr_tg    1199    -407.      7.09
## 3         1 Pck1ff 4hr_tg    2124   -2679.      7.66
## 4         2 Pck1ff 0hr_tg     63       1.45      4.14
## 5         2 Pck1ff 2hr_tg    742   -864.      6.61
## 6         2 Pck1ff 4hr_tg   6577   1774.      8.79
## 7         4 Pck1ff 0hr_tg     45    -16.5      3.81
## 8         4 Pck1ff 2hr_tg    955   -651.      6.86
## 9         4 Pck1ff 4hr_tg   3436  -1367.      8.14
## 10        5 Pck1ff 0hr_tg     33    -28.5      3.50
## # i 125 more rows

# Assuming your data frame is named 'your_data'
# and the columns are named 'dependent_variable', 'factor1', 'factor2'
# NA values removed during pivot_longer() above

# Perform two-way ANOVA
#model <- aov(dependent_variable ~ factor1 * factor2, data = your_data)

# Summarize the ANOVA results
#summary(model)

tg_ANOVA_LOGtf <- aov(TG_conc_log ~ time * genotype, data =
tg_tolerance_wide)

summary(tg_ANOVA_LOGtf)

##           Df Sum Sq Mean Sq  F value Pr(>F)
## time           2  466.1  233.07 1350.230 <2e-16 ***
## genotype       1    1.0    1.00   5.784 0.0176 *
## time:genotype   2    0.2    0.12   0.719 0.4890
## Residuals     129   22.3    0.17
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# Shapiro-Wilk test for normality
tg_tolerance_wide$residuals_LOG <- residuals(tg_ANOVA_LOGtf)

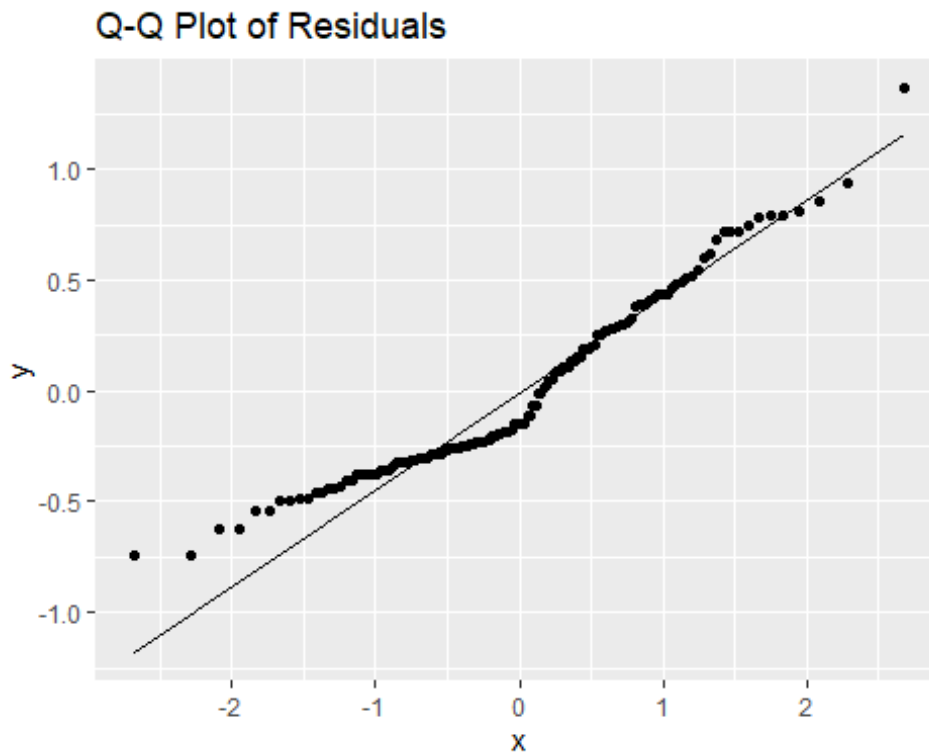
shapiro.test(tg_tolerance_wide$residuals_LOG)

```

```
##
## Shapiro-Wilk normality test
##
## data: tg_tolerance_wide$residuals_LOG
## W = 0.94578, p-value = 3.911e-05

# High p-value (e.g.,  $p > 0.05$ ): Data is likely normally distributed.
# Low p-value (e.g.,  $p \leq 0.05$ ): Data is likely NOT normally distributed.

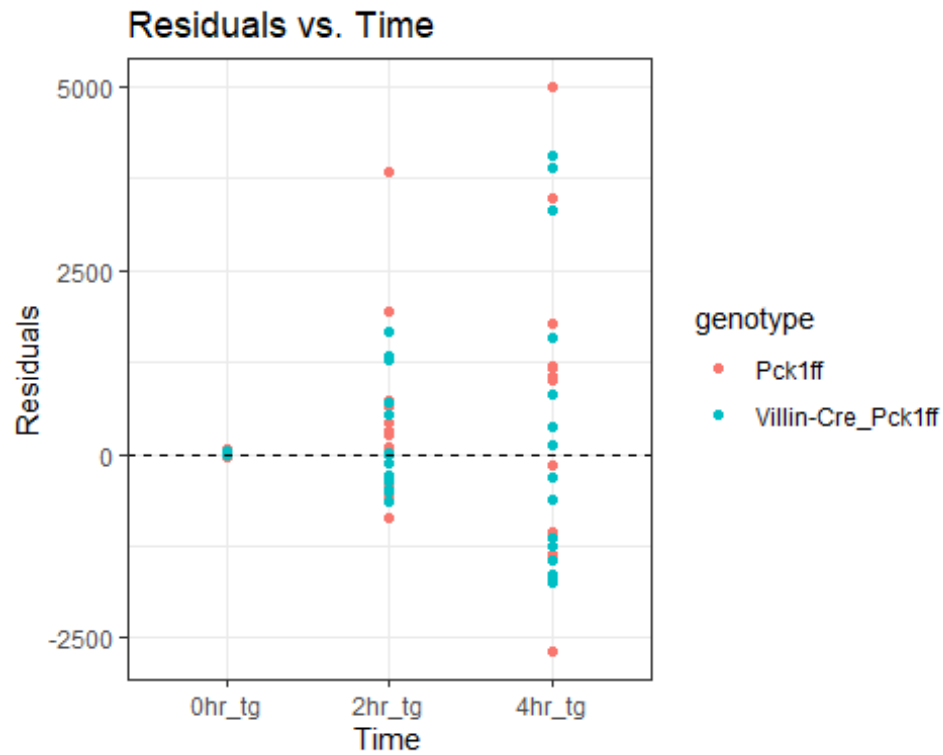
# Check for Normality
# Visual inspection with QQ-plot
ggplot(tg_tolerance_wide, aes(sample = residuals_LOG)) +
  stat_qq() +
  stat_qq_line() +
  ggtitle("Q-Q Plot of Residuals")
```



#Homoscedasticity 1/1:

```
ggplot(tg_tolerance_wide, aes(x = time, y = residuals, color = genotype)) +
  geom_point() +
  geom_hline(yintercept = 0, linetype = "dashed") +
  geom_smooth(method = "loess", se = FALSE, color = "blue") +
  labs(title = "Residuals vs. Time", x = "Time", y = "Residuals") +
  theme_bw()

## `geom_smooth()` using formula = 'y ~ x'
```



```
leveneTest(residuals(tg_ANOVA_LOGtf) ~ interaction(time, genotype), data =
tg_tolerance_wide) # interaction term for two-way
```

```
## Levene's Test for Homogeneity of Variance (center = median)
```

```
##      Df F value Pr(>F)
```

```
## group  5  0.2804 0.9232
```

```
##      129
```

```
# p<0.05; unequal variance
```

```
# p>0.05; acceptable variance
```