Glucose: pivot - mutate - ifelse

Introduction

There are different ways to organise a dataset. Statisticians almost always prefer the *long* format, while in other circumstances a wide format can make an overview easier. The long format contain one row for each observation, i.e. if observations are made at 4 time points for 14 horses this will lead to 4*14 rows of data. In the wide format instead each row represents one horse and the 4 measurements made on the same horse are shown in different columns and with different column names.

Plots and data manipulations, like producing new variables and sorting datasets is also part of the exercise.

Before you start, make sure you have installed and loaded the package tidyverse package.

```
library(tidyverse)
```

Read the glucose data set

This dataset contains glucose levels of Icelandic horses¹. Observations are made for 14 horses that get one of three different diets. Glucose and insulin levels are measured at a baseline, after 6, 12 and 24 hours. The dataset is in wide format.

```
glucose <- read_csv("Glucose.csv")</pre>
```

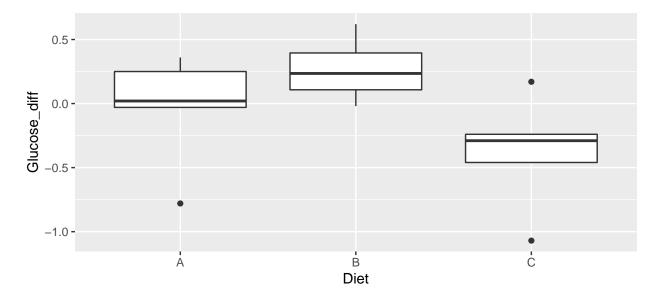
Use mutate to create new variables

We are interesting in investiging changes from the baseline glucose level at time point 24 hours. We compute both the glucose difference (Glucose_diff) and the ratio between measurement at time point 24 and the baseline (\text{Glucose_ratio}).

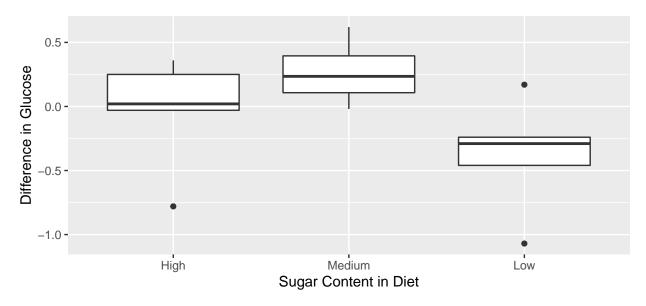
Plot your data

A boxplot can illustrate how the difference of glucose levels differs with diet.

 $^{^{1}}$ This is a simulated dataset based on a realistic experiment setup



The plot can be improved by setting more informative names on the x- and y-axes using labs for the x-and y-axis and \texttt{scale_x_discrete} to name the three diets (categories on the x-axis):



Exercise

- 1. Make boxplots for the glucose ratio computed above.
- 2. Compute differences between time point 12 and the baseline for insulin and make boxplots for this variable

Use gather to bring your dataset a long format

Gather all observation into one column with numeric data. We keep ID and Diet for each row and use the remaining column headers to produce a *key* for each row.

```
glucose_tidy <- glucose %>%
  pivot_longer(cols=c(Glucose_baseline: Insulin_24))
```

This means that observations can now be found in one column, while the *key* indicates which variable value is given. See the first entries of the data set:

head(glucose_tidy)

```
## # A tibble: 6 x 4
        ID Diet name
##
                                    value
##
     <dbl> <chr> <chr>
                                    <dbl>
## 1
         1 A
                  Glucose_baseline
                                    4.62
## 2
                  Glucose_6
                                     6.16
         1 A
## 3
                  Glucose_12
                                     4.6
         1 A
## 4
         1 A
                  Glucose_24
                                     4.87
## 5
         1 A
                  Insulin_baseline 61.5
## 6
         1 A
                  Insulin_6
                                    74.5
```

In the next step we separate the key into the variable name and the time point. Observe that you need to have names that are easy to split and that these names originate from the column names in the original data set. In this case the names are separated by an underscore (_) which is passed to the program with the $sep = " \setminus _"$ statement.

head(glucose_tidy)

```
## # A tibble: 6 x 5
##
        ID Diet variable time
                                     value
##
     <dbl> <chr> <chr>
                           <chr>
                                     <dbl>
## 1
         1 A
                  Glucose
                           baseline
                                      4.62
## 2
         1 A
                  Glucose
                           6
                                      6.16
## 3
         1 A
                  Glucose
                           12
                                      4.6
## 4
         1 A
                  Glucose
                           24
                                      4.87
## 5
         1 A
                  Insulin baseline 61.5
                                     74.5
## 6
         1 A
                  Insulin
```

Usually we do not want to have Insulin and Glucose in the same column. To split it into two columns according to variable name the statement spread can be used:

The fist observations in the new dataset are seen below. We have now 5 columns: ID, Diet, time, Glucose, Insulin.

head(glucose_tidy)

```
## # A tibble: 6 x 5
##
        ID Diet time
                            Glucose Insulin
##
     <dbl> <chr> <chr>
                              <dbl>
                                       <dbl>
## 1
         1 A
                  baseline
                               4.62
                                        61.5
## 2
         1 A
                  6
                                6.16
                                        74.5
## 3
         1 A
                  12
                               4.6
                                       117.
## 4
         1 A
                  24
                               4.87
                                        83.7
## 5
         2 A
                  baseline
                                4.88
                                        49.1
## 6
         2 A
                                4.1
                                        70.2
```

If we want to rename the time point now called baseline to 0 we can use the mutate statement and change the variable time if (and only if) is is equal to baseline. The ifelse statement gives first the condition (if time == "baseline"), then value value that is given if the condition is true (0) and then the value if the condition is not true (time, i.e. the other values of time).

```
glucose2 <- glucose_tidy %>%
  mutate(time = ifelse(time == "baseline", 0, time))
```

To inspect the dataset we can use View.

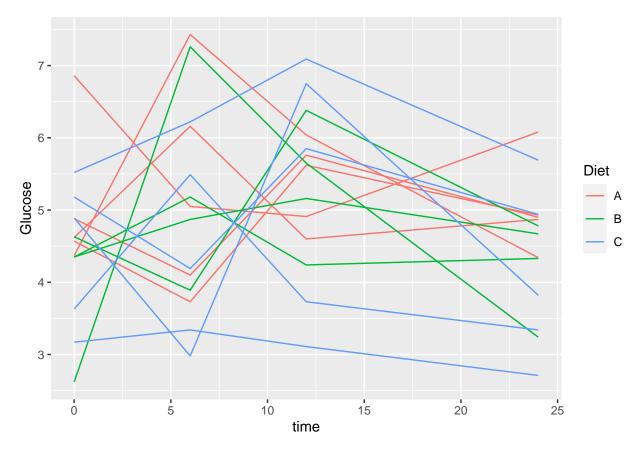
```
View(glucose2)
```

In this case time is a character and ID is numeric. Reasonably, ID should be a factor and time could be either a factor or a numeric variable. Let's set it to a numeric variable here, even if it probably would be best to use as a factor in a statistical analysis.

By using arrange we can bring the data in a logical order, e.g. sorted by Treatment (Diet), Animal ID, and time in the order 0, 6, 12, and 24.

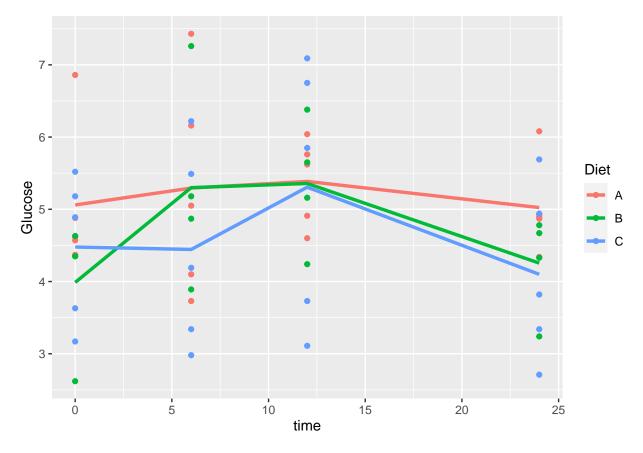
Plot glucose data

A line plot can be used to illustrate the time series for each horse. To get individual lines for each horse use the <code>group=ID</code> statement. Different colors are used for the three different diets.



It might also be good to include a mean line that illustrates how the mean for each diet changes over time. This can be obtained by stat_summary. Line width (lwd) is here set to 1.2 to get a thicker line for this mean.

Warning: `fun.y` is deprecated. Use `fun` instead.



Exercise

3. Make a similar plot for the insulin data.

Solutions to exercices

1.

```
Glucose_change %>%
  ggplot(aes(x = Diet,
            y = Glucose_ratio)) +
  geom_boxplot() +
  labs(y = "Glucose ratio",
       x = "Sugar Content in Diet") +
  scale_x_discrete(labels = c("High", "Medium", "Low"))
Glucose_change <- glucose %>%
  mutate(Glucose_diff = Glucose_24 - Glucose_baseline,
         Glucose_ratio = Glucose_24 / Glucose_baseline,
         Insulin_diff = Insulin_12 - Insulin_baseline)
Glucose_change %>%
  ggplot(aes(x = Diet,
            y = Insulin_diff)) +
  geom_boxplot() +
  labs(y = "Difference in Insulin",
       x = "Sugar Content in Diet") +
  scale_x_discrete(labels = c("High", "Medium", "Low"))
  3.
glucose2 %>%
  ggplot(aes(x = time,
             y = Insulin,
             group = ID,
             color = Diet)) +
  geom_line()
glucose2 %>%
  ggplot(aes(x = time,
             y = Insulin,
             group = ID,
             color = Diet)) +
  geom_point() +
  stat_summary(aes(group = Diet),
               fun = mean,
               geom = "line",
               lwd = 1.2)
```