Reading, Summarizing, Visualizing Data

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In this note, we'll read in a data set from a csv file. We'll then get some summary statistics on the variables, and then visualize the data.

Make sure the necessary R packages are loaded

We'll need three packages:

- readr for reading in the csv file
- dplyr for doing summaries on the data
- ggplot2 for visualizing and plotting the data

These and any other package can be loaded by using the library() command like this:

```
library(readr)
library(dplyr)
library(ggplot2)
```

Also, if these packages are not yet installed, you must install them. They can all be installed by installing the package-of-packages called tidyverse.

Read in the data

The data file we will be using is called fake_data_01.csv. If this data file is in your so-called "working directory", then you can read it in, and save it by the name Df, with the following command:

```
Df <- read_csv('fake_data_01.csv')</pre>
```

You can also effectively run the above command by going to the Environment tab in Rstudio (upper right panel), and choose "Import Dataset" and then "From Text (readr)..".

Let's look at the data

• First, we can just type the name of the data-frame. By default, you'll see just the first 10 rows, and some general info about the size of the data set.

 Df

```
## # A tibble: 100 x 3
##
                 iq test_score
      sex
##
      <chr>
             <int>
                          <dbl>
##
                           14.1
    1 male
                 74
##
    2 female
                 83
                           16.0
##
    3 female
                 88
                           15.6
    4 male
                105
                           11.8
##
##
                 92
                           13.5
   5 male
                           13.4
    6 male
                 92
                           12.7
##
    7 male
                102
    8 female
                 80
                           14.9
```

```
## 9 male 102 14.1
## 10 female 114 15.7
## # ... with 90 more rows
```

• And then we can get another look via glimpse:

<int> 74, 83, 88, 105, 92, 92, 102, 80, 102, 114, 107, 80...

Summarize the data

• We can always get a general summary of a data frame via the summary command:

\$ test_score <dbl> 14.14571, 16.01274, 15.64022, 11.82935, 13.48280, 1...

summary(Df)

\$ iq

```
##
                                            test_score
        sex
                               iq
                                : 60.0
                                                  :11.49
##
    Length: 100
                        Min.
                                          Min.
                         1st Qu.: 91.0
                                          1st Qu.:13.26
##
    Class : character
    Mode :character
                        Median :102.0
                                          Median :14.63
##
                                :100.4
                                                  :14.50
                         Mean
                                          Mean
##
                        3rd Qu.:108.2
                                          3rd Qu.:15.65
##
                                :138.0
                        Max.
                                          Max.
                                                  :18.40
```

• But we can get more specific information using the summarize (also known as summarise) command. With this command, we can ask for many different summary statistics, on as many or as few variables as we are interested in. Let's say we want to know the average iq, average test_score, standard deviation of iq and the variance of test_score. We'll also get the number of data points we have overall, using the n() function. Then we do the following:

```
summarize(Df,
          average_iq = mean(iq),
          average_score = mean(test_score),
          iq_stdev = sd(iq),
          score_var = var(test_score),
          n = n()
)
## # A tibble: 1 x 5
##
     average_iq average_score iq_stdev score_var
##
          <dbl>
                         <dbl>
                                   <dbl>
                                              <dbl> <int>
           100.
                          14.5
                                    14.5
## 1
                                              2.30
                                                      100
```

Summarize by a group

We can get summary statistics per each group by doing a group_by operation first. So let's say we wanted to get, as before, the average iq, average test_score, standard deviation of iq and the variance of test_score, but this time, we want these summary statistics separately for the males and females, then we do this:

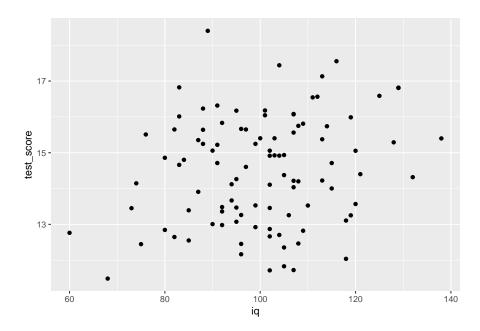


Figure 1: A scatterplot of each participant's test score as a function of their IQ score.

```
iq_stdev = sd(iq),
          score_var = var(test_score),
          n = n()
## # A tibble: 2 x 6
##
     sex
            average_iq average_score iq_stdev score_var
                                                                n
                  <dbl>
##
     <chr>>
                                 <dbl>
                                           <dbl>
                                                     <dbl> <int>
## 1 female
                                  15.4
                                            15.6
                                                      1.46
                  101.
                                                               51
## 2 male
                   99.6
                                  13.6
                                            13.4
                                                      1.48
                                                               49
```

Visualization

• First, let's make a scatterplot. We can do so using the following ggplot code:

The resulting figure is shown in Figure 1.

• We can add a line of best fit by adding stat_smooth(method='lm') like this:

This figure is shown in Figure 2.

• We can colour the points according to whether they are male of female by adding col = sex within the aes() command like this:

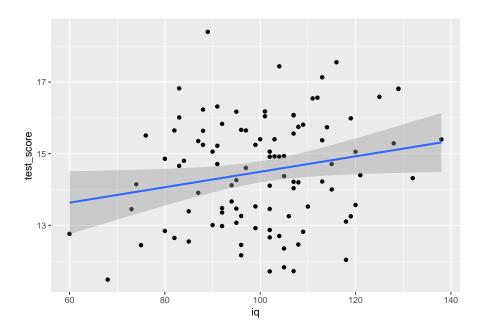


Figure 2: A scatterplot of each participant's test score as a function of their IQ score. Also shown is line of best fit to the data.

This is shown in Figure 3.

• And we can make a line of best fit for the males and females separately:

This is shown in Figure 4.

• We can change the style in many ways. Here, is one possibility:

This is shown in Figure 5.

Tukey boxplots

fig_cap <- "A Tukey boxplot showing the distribution of the test score variables for males and females

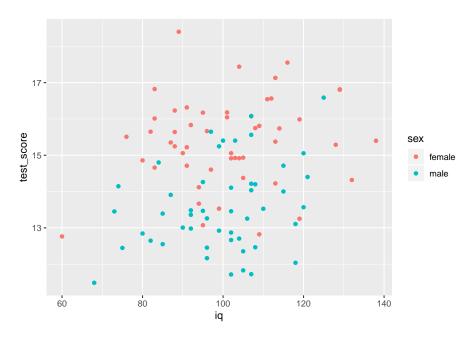


Figure 3: A scatterplot of each participant's test score as a function of their IQ score. The data points from males and females are identified by different colours.

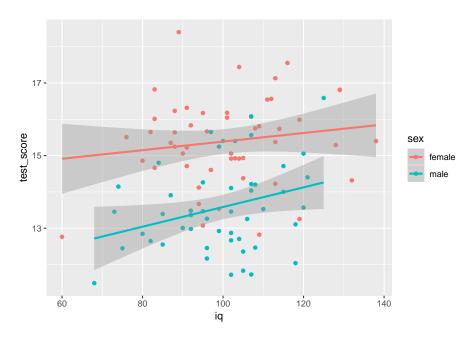


Figure 4: A scatterplot of each participant's test score as a function of their IQ score. The data points from males and females are identified by different colours. The lines of best fit for the data from males and females separately is also shown.

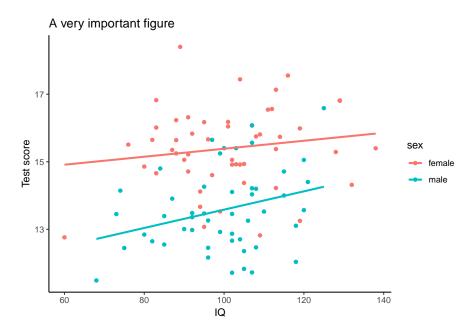


Figure 5: A scatterplot of each participant's test score as a function of their IQ score. The data points from males and females are identified by different colours. The lines of best fit for the data from males and females separately is also shown. In this figure, we change the plotting style, add new axis labels, and add a figure title.

To visualize distributions of data, a venerable old method is the Tukey boxplot. Here, we will visualize the distribution of test_score for the males and females separately.

The resulting figure is shown in Figure 6. By default, boxplots show, in the box part of the plot, the lower quartile, median, and upper quartile values. The whiskers extend from the upper and lower quartiles to the furtherest points that are not outliers, where outliers are defined as being within 1.5 times the inter-quartile range from the box. We can, in addition, provide all of the points by adding a so-called jitter plot on top of the boxplot.

The resulting figure, Figure 7, is a bit of unsightly. But we can make it more presentable very easily.

The resulting figure in Figure 8.

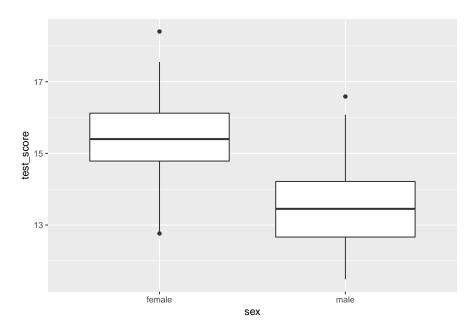


Figure 6: A Tukey boxplot showing the distribution of the test score variables for males and females separately.

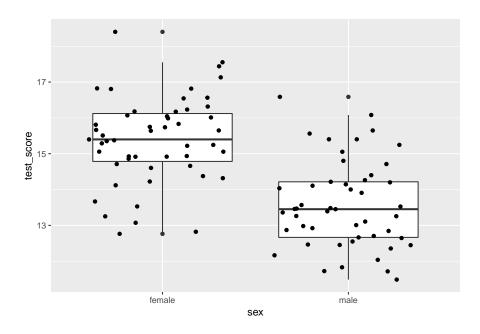


Figure 7: A Tukey boxplot showing the distribution of the test score variables for males and females separately. In adddition, a jitter plot of all points is shown.

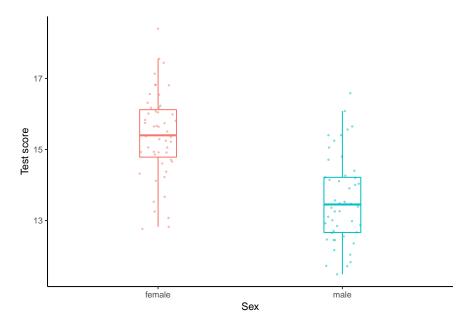


Figure 8: A Tukey boxplot showing the distribution of the test score variables for males and females separately. In adddition, a jitter plot of all points is shown. This version is just a bit nicer to look at than the previous one.