

Bad data & Outliers

true

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```
pacman::p_load(  
  broom,  
  conflicted,  
  here,  
  janitor,  
  readxl,  
  tidyverse  
)
```

There are two download links:

- Download the **original** excel file [here](#).
- Download the **formatted** excel file [here](#).

1 Data

Imagine that this dataset was obtained by you. You spent an entire day walking around the campus of a university and asked a total of 29 people for things like how old they are (**Ages**) and you also tested how well they could see on a scale of 1-10 (**Vision**).

1.1 Import

Assuming you are working in a [R-project](#), save the formatted file somewhere within the project directory. I have saved it within a sub folder called **data** so that the relative path to my file is `data/vision_fixed.xls`.

```
path <- here("data", "vision_fixed.xls")
dat <- read_excel(path)
```

```
dat
```

```
# A tibble: 29 x 9
  Person      Ages Gender `Civil state` Height Profession Vision Dista~1 PercD~2
  <chr>      <dbl> <chr>   <chr>          <dbl> <chr>      <dbl>   <dbl>   <dbl>
1 Andrés      25 M     S              180 Student      10     1.5     15
2 Anja        29 F     S              168 Professio~  10     4.5     45
3 Armando    31 M     S              169 Professio~   9     4.5     50
4 Carlos     25 M     M              185 Professio~   8     6       75
5 Cristina   23 F     <NA>          170 Student      10     3       30
6 Delfa      39 F     M              158 Professio~   6     4.5     75
7 Eduardo    28 M     S              166 Professio~   8     4.5    56.2
8 Enrique    NA <NA>   <NA>           NA Professio~  NA     6      NA
9 Fanny      25 F     M              164 Student       9     3     33.3
10 Francisco  46 M     M              168 Professio~   8     4.5    56.2
# ... with 19 more rows, and abbreviated variable names 1: Distance,
# 2: PercDist
```

1.2 Goal

Very much like in the previous chapter, our goal is to look at the relationship of two numeric variables: **Ages** and **Vision**. What is new about this data is, that it (i) has missing values and (ii) has a potential outlier.

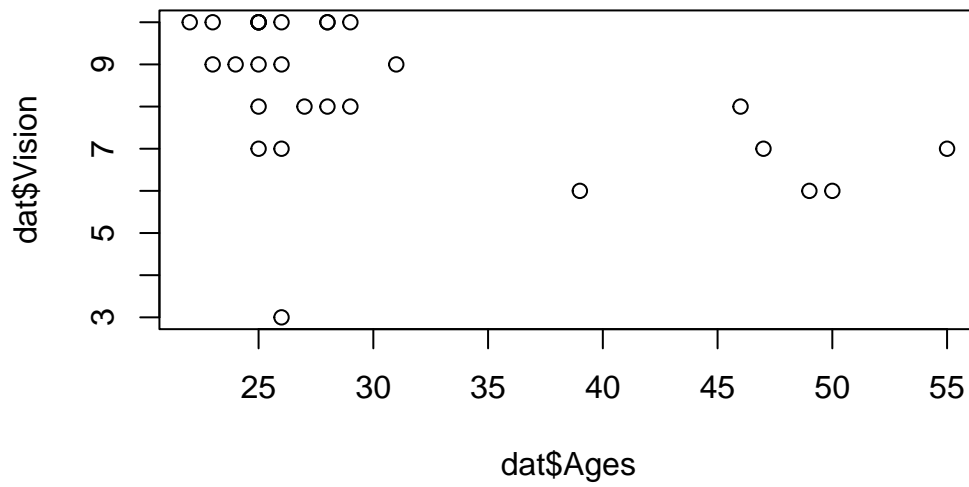
1.3 Exploring

To quickly get a first feeling for this dataset, we can use `summary()` and draw a plot via `plot()` or `ggplot()`.

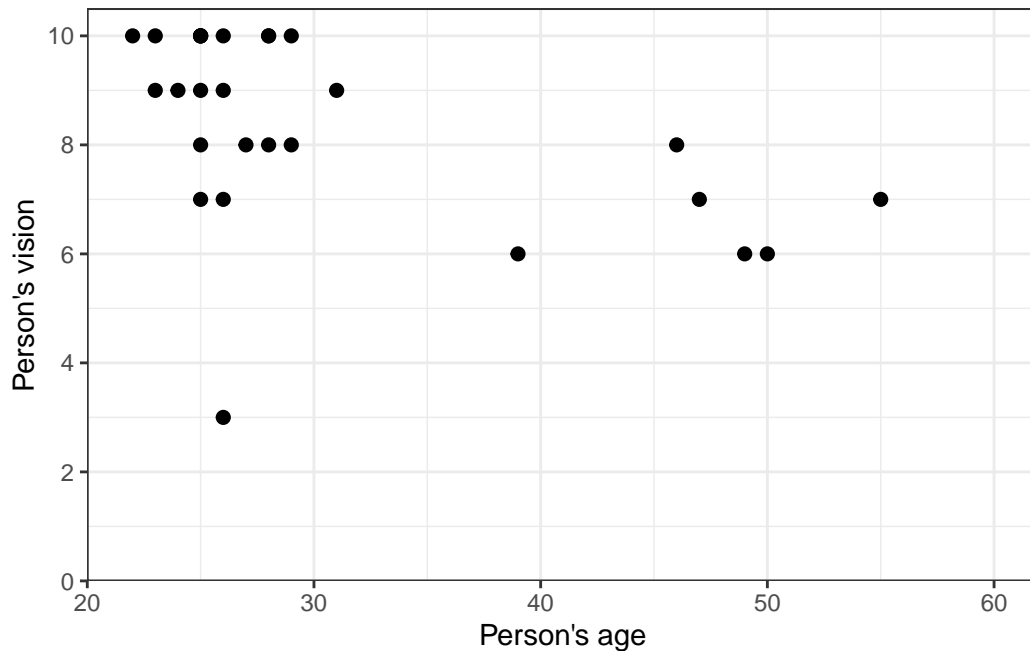
```
summary(dat)
```

Person	Ages	Gender	Civil state
Length:29	Min. :22.00	Length:29	Length:29
Class :character	1st Qu.:25.00	Class :character	Class :character
Mode :character	Median :26.00	Mode :character	Mode :character
	Mean :30.61		
	3rd Qu.:29.50		
	Max. :55.00		
	NA's :1		
Height	Profession	Vision	Distance
Min. :145.0	Length:29	Min. : 3.000	Min. :1.500
1st Qu.:164.8	Class :character	1st Qu.: 7.000	1st Qu.:1.500
Median :168.0	Mode :character	Median : 9.000	Median :3.000
Mean :168.2		Mean : 8.357	Mean :3.466
3rd Qu.:172.8		3rd Qu.:10.000	3rd Qu.:4.500
Max. :190.0		Max. :10.000	Max. :6.000
NA's :1		NA's :1	
PercDist			
Min. : 15.00			
1st Qu.: 20.24			
Median : 40.18			
Mean : 45.45			
3rd Qu.: 57.19			
Max. :150.00			
NA's :1			

```
plot(y = dat$Vision, x = dat$Ages)
```



```
ggplot(data = dat) +
  aes(x = Ages, y = Vision) +
  geom_point(size = 2) +
  scale_x_continuous(
    name = "Person's age",
    limits = c(20, 60),
    expand = expansion(mult = c(0, 0.05))
  ) +
  scale_y_continuous(
    name = "Person's vision",
    limits = c(0, NA),
    breaks = seq(0, 10, 2),
    expand = expansion(mult = c(0, 0.05))
  ) +
  theme_bw()
```



Apparently, most people are in their 20s and can see quite well, however some people are older and they tend to have a vision that's a little worse.

2 Correlation & Regression

Let's estimate the correlation and simple linear regression and look at the results in a tidy format:

```
cor <- cor.test(dat$Vision, dat$Ages)
tidy(cor)
```

```
# A tibble: 1 x 8
  estimate statistic p.value parameter conf.low conf.high method      alter~1
  <dbl>      <dbl>   <dbl>      <int>   <dbl>    <dbl> <chr>      <chr>
1  -0.497      -2.92 0.00709        26  -0.734    -0.153 Pearson's pro~ two.si~
# ... with abbreviated variable name 1: alternative
```

```
reg <- lm(Vision ~ Ages, data = dat)
tidy(reg)
```

```
# A tibble: 2 x 5
  term      estimate std.error statistic  p.value
<chr>      <dbl>      <dbl>      <dbl>    <dbl>
1 (Intercept)  11.1        0.996        11.2 1.97e-11
2 Ages        -0.0910    0.0311        -2.92 7.09e- 3
```

Thus, we have a negative, moderate correlation of -0.497 and for the regression we have $Vision = 11.14 + -0.09 \text{ Ages}$.