Data Science for Socioeconomists

Excercise 6 - Data visualization pt. 1

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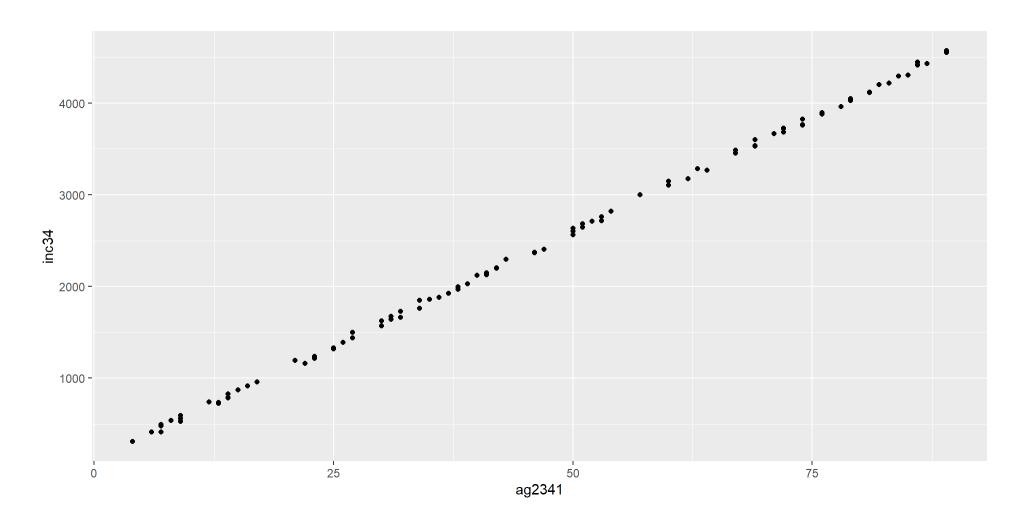
- We have created a link-list with all things helpful this can be found under "learning materials" at OpenOlat
- The code of the lecture-, exercise- and solution slides can be copied! This way you can adapt it to your needs and don't need to start from scratch.

- 1. Axis labeling
- 2. scales
- 3. legends
- 4. readability
- 5. color usage
- 6. accessibility
- 7. Tasks for today

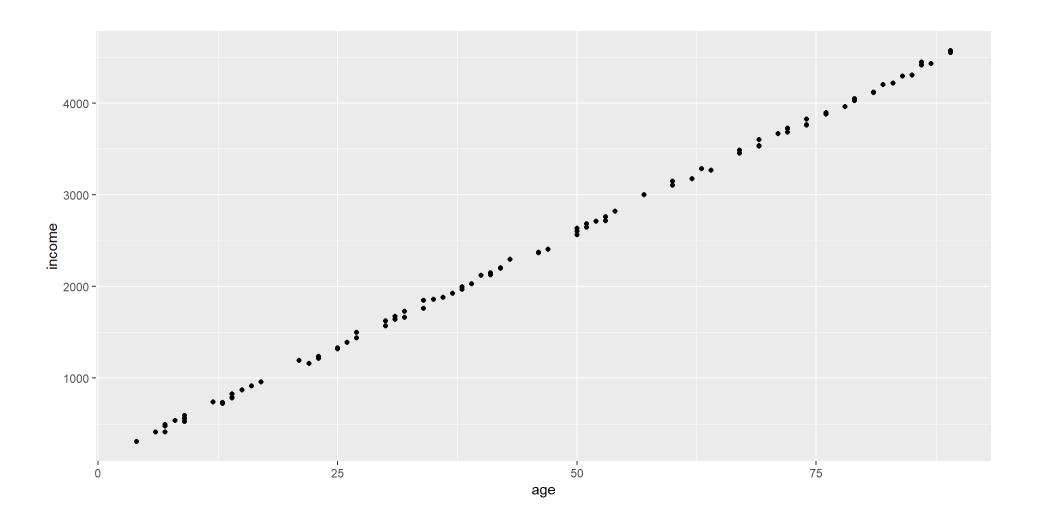
- We can look at:
 - labeling the axis
 - decision about units
 - aspect ratio (wide enhances changes on the x-axis, narrow and tall enhances changes on the y-axis)
 - grid spacings

We get the axis title of the variable we inserted - ugly names create ugly axis titles

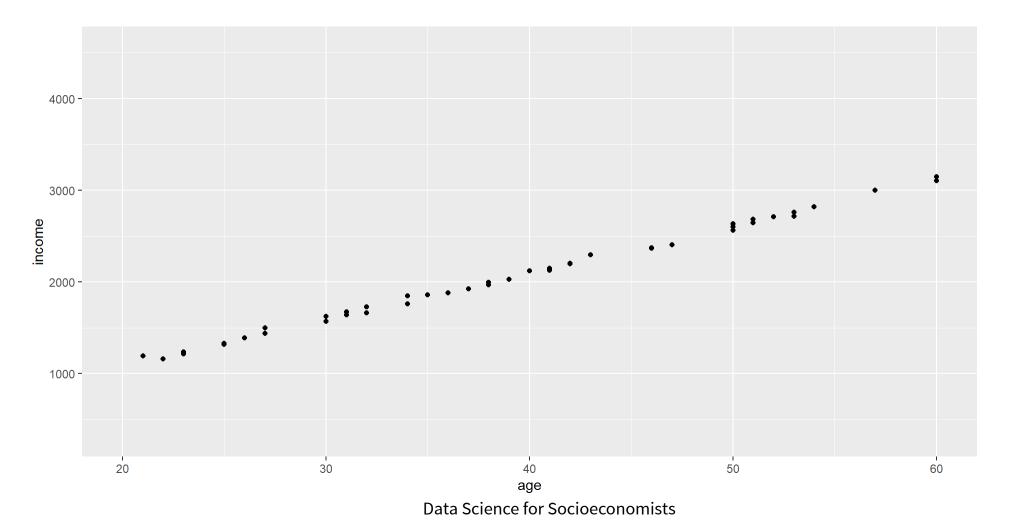
```
1 library(sozoekds)
2 library(ggplot2)
3 test_dataset <- testdata
4 test_dataset$ag2341 <- test_dataset$age
5 test_dataset$inc34 <- test_dataset$income
6 ggplot(data=test_dataset) +
7 geom_point(aes(ag2341,inc34))</pre>
```



```
1 ggplot(data=test_dataset) +
2 geom_point(aes(ag2341,inc34)) +
3 xlab("age") +
4 ylab ("income")
```

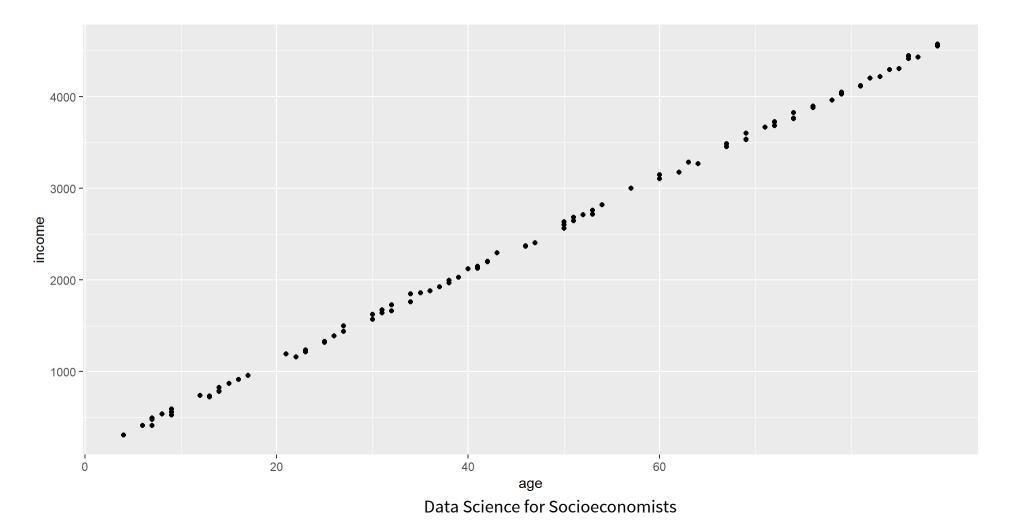


changing the range of values that are plotted

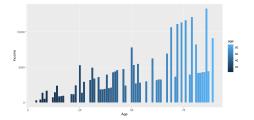


setting "tick marks"

```
1  ggplot(data=test_dataset) +
2   geom_point(aes(ag2341,inc34)) +
3   xlab("age") +
4   ylab ("income") +
5  scale_x_continuous(breaks=c(0, 20, 40, 60))
```



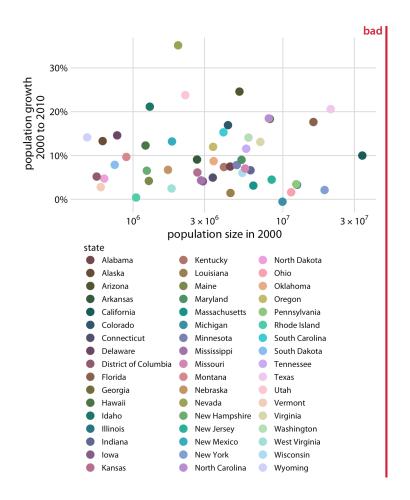
```
1 ggplot(data = test_dataset, aes(x = age, y = income, fill = age)) +
2   geom_bar(stat = "identity") +
3   xlab("Age") +
4   ylab("Income")
```

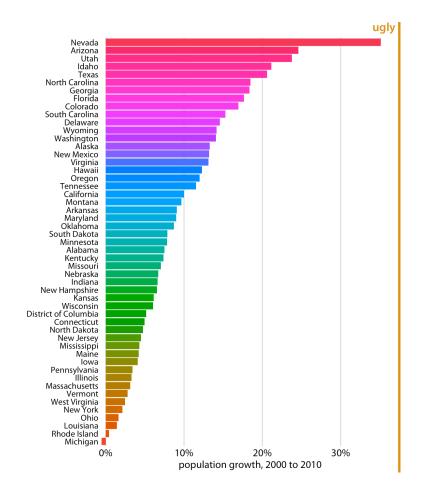


we can also:

- change the order of the elements in our legend
- change the name of the categories in the legend
- change the position of the legend.... and a lot more

What's bad about these visualizations? What can we do better?





via: https://clauswilke.com/dataviz /pitfalls_of_color_use_files/figurehtml/popgrowth-vs-popsize-colored-1.png via: https://clauswilke.com/dataviz/pitfalls_of_color_use_files/figure-html/popgrowth-US-rainbow-1.png

colors

- Fewer colors enable the brain to process the information more clearly
- There are three main types of color palettes used in the world of data visualizations:
 - Qualitative palette each color is distinct from the others
 - Sequential palette a single color in a variety of saturations or a gradient
 - Diverging palette color variables sit on a spectrum, such as cold to hot
 - What can we change in the graphics we saw before?

colors need to be easy to distinguish - What about color-blindness?

- 1. deutanomaly: green-blind
- 2. protanomaly: red-blind
- 3. tritanomaly: blue-blind

⇒ check acessability via a website or via an R-package colorblindr

1. Revisit last weeks plots - how can you transform them with the principles we learned today?

Choose **one** of the two data sets and do the exercises regarding this set.

 Create a new variable called Avg_Rooms that displays the average amount of rooms in each household in a certain neighborhood (block)

Hint: for one observation there might be 20 households, and 50 rooms in total - this leads to an average of 2.5 rooms per household

- Plot Avg_Rooms against Median_Income interpret the result
- 3. Eliminate values in Avg_Rooms that are greater then 10 rooms from your data set then plot task 2 again
- 4. as before: Explain your decision of the type of visualization + Interpret the results

Data Science for Socioeconomists

- 1. Create a new variable called full_score that is the mean score of all three testscores (Math, Reading, Writing)
- 2. Calculate grades from the full_score using the american system (hint here)
- 3. Plot the distribution of grades by number of study hours per week

Hint: create a factor variable from WklyStudyHours

Hint: You could do this in one plot with multiple lines or in more than one graphic where all the plots are printed in one frame (https://intro2r.com/mult_graphs.html)

4. as before: Explain your decision of the type of visualization + Interpret the results

Starting next week my colleague Victoria Hünewaldt will take over and start with statistical learning and machine learning. We'll meet again in January!