# Datenvisualiserung 1

## Distributions

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## Wiederholung

Last week we...

- installed R and RStudio
- created our first R script
- did simple arithmetic with objects and vectors

## Wiederholung

```
x \leftarrow c(1,2,3)

y \leftarrow sum(1,2,3)
```

- what do the vectors **x** and **y** contain?
- The object x contains 1, 2, 3.
- The object y contains 6.

## Heutige Ziele

Today we will learn...

- what dataframes are
- the difference between categorical and continuous data
- how to produce plots with ggplot
- choose the right plot for our data

#### Lust auf mehr?

- Chapter 2 (Data Visualisation) in Wickham et al. (2023), up until section 2.4
- Chapter 3 (Data Visualisation) in Nordmann & DeBruine (2022)

## Vorbereitung

In your RProject folder...

- create a new folder called moodle
  - download the Moodle materials from today and save them there
- create a new folder in notes called 02-datenviz1
- open a new .R script
  - save it in the new folder

## 0.0.1 Packages

- Pakete (installiert und) ladt
  - tidyverse
  - languageR
  - ggthemes
  - patchwork

```
# in the CONSOLE: install packages if needed
install.packages("tidyverse")
install.packages("languageR")
install.packages("ggthemes") # for customising our plots
install.packages("patchwork") # plot layouts
```

```
# Pakete laden
library(tidyverse)
library(languageR)
library(ggthemes)
library(patchwork)
```

#### 1 Data frames

- data frames are a collection of variables, where
  - each variable is one column
  - each row is a single observation/data point
  - each cell in a row is linked
- data frames are just like spreadsheets, but are rectangular
- different words for data frames:
  - data frame
  - dataset
  - tibble (in the tidyverse)

## 1.1 Talking about datasets

- when we talk about our data, we use certan words to refer to different parts:
- a variable: a quantity, quality, or property you can measure
- a value: the state of a variable when you measure it
- an **observation**: set of measurements made under similar conditions
  - will contain several values each associated with a variable
  - an observation for a single variable is sometimes called a data point
- tabular data is a set of values, each associated with a variable and an observation
  - tabular data is *tidy* if each value is placed in its own *cell*, each variable in its own column, and each observation in its own row

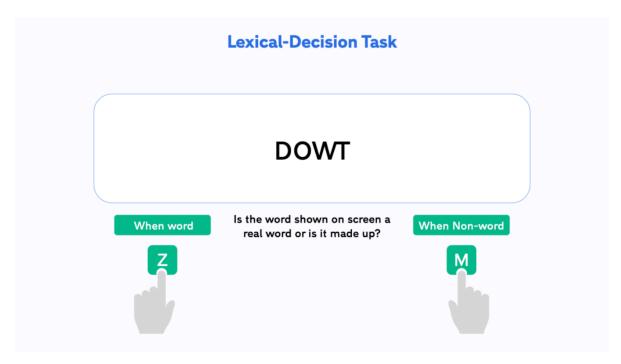
## 1.2 Categorical and continuous variables

- how we visual the distribution of a variable depends on what type of data it represents: categorical or numerical
- a variable is *categorical* if it can take a small set of values that can be grouped together
- e.g., old/young, short/tall, grammatical/ungrammatical, L1/L2-speaker
- a variable is *numerical* (i.e., quantitative) if it can take on a wide range of numerical values
  - and it would make sense to add, subtract, compute the mean, etc.
  - can be *continuous* (decimal points make sense, e.g., 1.5cm)

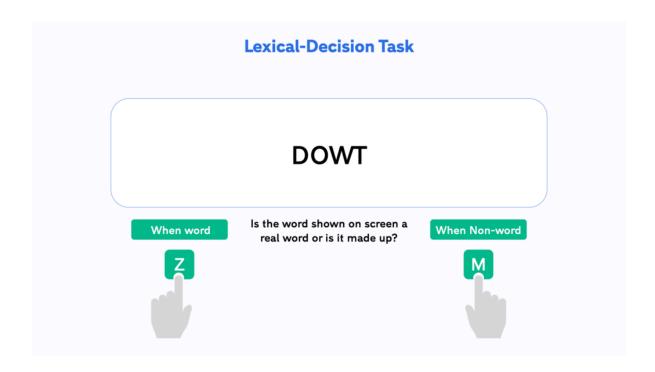
- or discrete (decimal points do not make sense, e.g., 1.5 children doesn't make sense)
- age, height, reaction times, format frequencies
- we produce different plots depending on what type of variables we want to visualise

## 2 Lexical Decision Task

- our first dataset contains data from a lexical decision task (LDT)
- in the LDT, participants press a button to indicate whether a word is a real word or a pseudoword



 $Figure~1:~Source:~https://www.testable.org/wp-content/uploads/2022/11/Lexical\_decision\_task-1024x576.png$ 



#### 2.1 LDT variables

- common variables collected in a lexical decision task experiment are:
  - reaction time
  - accuracy (correct/incorrect)
  - word category (e.g., real/pseudo, noun/verb)
  - word frequency
- additional variables that might be collected could be:
  - participant demographics (e.g., age, L1/L2, gender)

## 3 lexdec dataset

- languageR is a companion package for the textbook Baayen (2008)
  - contains linguistic datasets, e.g., lexdec
- the lexdec dataset contains data for a lexical decision task in English
  - we will be working with variables such as reaction times and accuracy

#### 3.1 lexdec variables

• a list of some of the variables is included in Table 1

Table 1: Data dictionary for df\_lexdec: Lexical decision latencies elicited from 21 subjects for 79 English concrete nouns, with variables linked to subject or word.

variable	description
Subject	a factor for the subjects
RT	a numeric vector for reaction times in milliseconds
Trial	a numeric vector for the rank of the trial in the experimental list.
Sex	a factor with levels F (female) and M (male).
${\bf Native Language}$	a factor with levels English and Other, distinguishing between native and nonnative speaker

#### 3.2 LDT research questions

- before we conduct an experiment, we have research questions that we want to answer with the data
  - today we'll address the following question:
    - \* do the reaction times differ between native and non-native speakers?

#### 3.3 Load the data

- our data is available in the lanaugeR package we've already loaded
  - to print the data, just type the name of the dataset and run it
- below we only see a few variables, but you should see more in your console

#### lexdec

PrevCorrect	PrevType	Correct	NativeLanguage	Sex	Trial	RT	Subject	
correct	word	correct	English	F	23	6.340359	A1	1
correct	nonword	correct	English	F	27	6.308098	A1	2
correct	nonword	correct	English	F	29	6.349139	A1	3
correct	word	correct	English	F	30	6.186209	A1	4
correct	nonword	correct	English	F	32	6.025866	A1	5
correct	word	correct	English	F	33	6.180017	A1	6

• how many variables do we have? observations?

### 3.3.1 Save data as an object

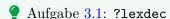
- to save the data in our Environment, we have to assign it a name
  - let's call it df\_lexdec, which means "data frame lexical decision"

#### df\_lexdec <- lexdec</pre>

- now we see it in our Enrivonment
  - double-click on it to view it in the Editor pane

#### 3.4 Relevant variables

- Among the variables we have are:
  - 1. Subject: participant ID
  - 2. RT: logged reaction times
  - 3. NativeLanguage: the native language of the participant
  - 4. Word: what word was presented
  - 5. Class: if the word was animal or plant

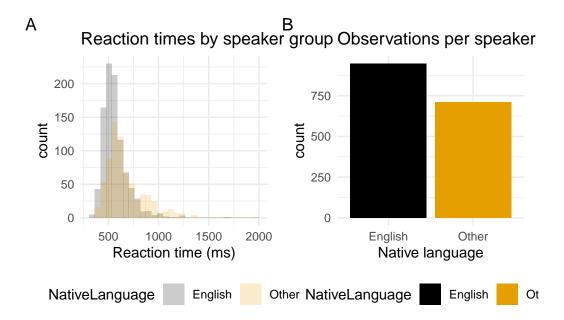


#### Example 3.1.

Find out what the other variables represent by running ?lexdec in the console.

## 3.5 Ultimate goal

- our ultimate goal today is to produce the following visualisation of the data
  - the plot shows the distribution (count) of reaction times and native language of the participants



## 4 Creating plots with ggplot2

- the tidyverse is a collection of packages that facilitate data tidying and visualisation
   when we load the tidyverse, this collection of packages is automatically loaded
- the ggplot2 package is a tidyverse package that builds plots in layers

## ggplot2 Schichten

## 4.1 Layer 1: empty canvas

• the first layer with the function ggplot() is like an empty canvas

```
ggplot(data = df_lexdec)
```

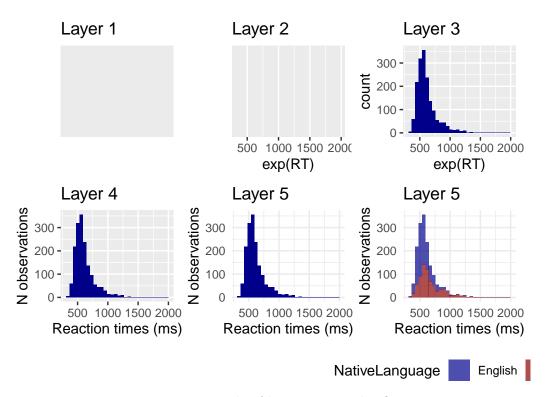
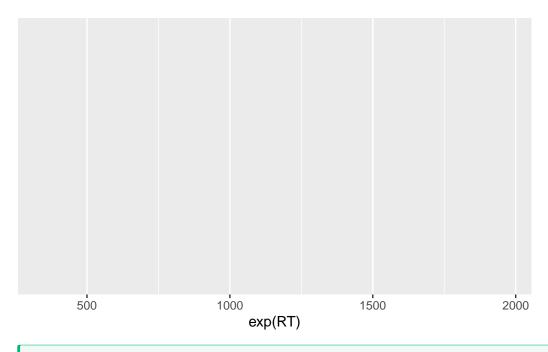


Figure 2: Example of layers in a ggplot figure

## 4.2 Layer 2: mapping aesthetics

- next we tell ggplot() how to visually represent our variables
  - we add the + to the end of our line of code, and on a new line of code use the function aes() to define our aethetics
- our first aesthetic maps reaction times (RT) on the x-axis (the bottom of the plot)
  - we wrap the logged RT in the exp() function to get RTs in milliseconds (for reasons we won't discuss)

```
ggplot(data = df_lexdec) +
aes(x = exp(RT))
```



• Aufgabe 4.1: Mapping aesthetics

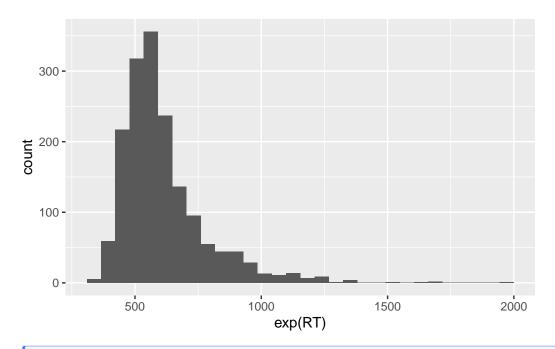
## Example 4.1.

1. Produce the plot so far

## 4.3 Layer 3: adding observations

- we don't see any observations (i.e., the bars) on the plot, why not?
  - we haven't told ggplot() how to visualise them
- we have to define a **geom**: the *geometrical* object that a plot uses to represent data
  - in ggplot2, geom functions start with geom\_
  - we often describe plots in terms of types of geoms they use, e.g., bar charts use bar geoms (geom\_bar()), line charts use line geoms (geom\_line()), scatterplots use a point geom (geom\_point()), etc.
- let's create our histogram using the geom geom\_histogram()

```
ggplot(data = df_lexdec) +
aes(x = exp(RT)) +
geom_histogram()
```



## i Note

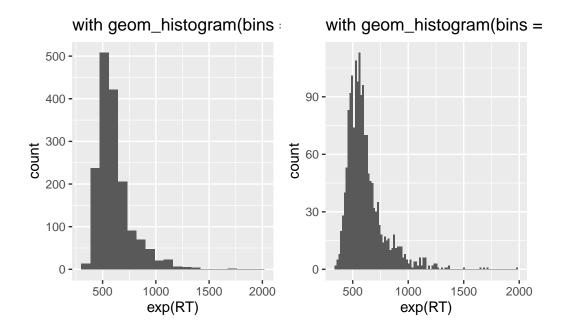
We got the following message when including geom\_point():

stat\_bin() using bins = 30. Pick better value with binwidth.

This is just telling us about the width of our bars: each bar represents a range of possible reaction time values + bins = 30 simply means there are 30 bars, we can change this have more or fewer bars by including e.g., bins = 20 or bins = 100 inside geom\_histogram()

```
ggplot(
  data = df_lexdec,
  mapping = aes(x = exp(RT))
) +
  labs(title = "with geom_histogram(bins = 20)") +
  geom_histogram(bins = 20) +

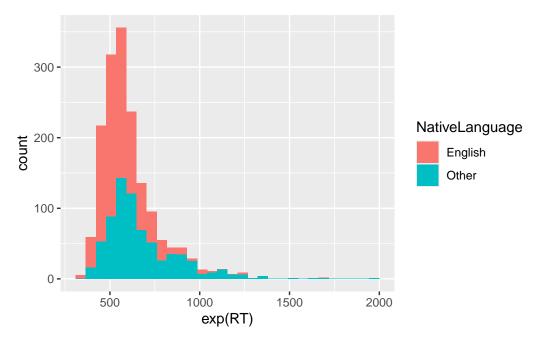
  ggplot(
  data = df_lexdec,
  mapping = aes(x = exp(RT))
) +
  labs(title = "with geom_histogram(bins = 100)") +
  geom_histogram(bins = 100)
```



## 4.4 Adding aesthetics

- seeing the distribution of reaction times in general is useful
  - but we usually want to compare groups
  - e.g., differences between native and non-native speakers, or between types of words
- we also have native language as a variable, how might we visualise this in our plot?

```
ggplot(
  data = df_lexdec,
  aes(x = exp(RT), fill = NativeLanguage)
) +
  geom_histogram()
```

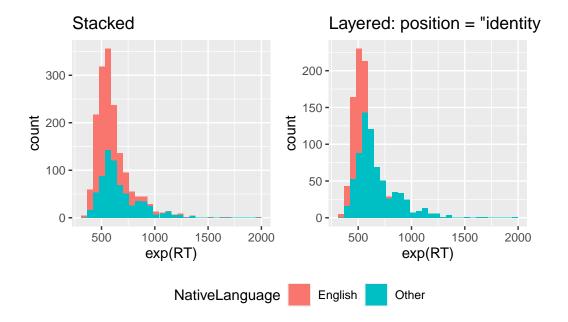


- we see the red bars and the blue bars, but is the blue histogram layered on top of the red?
  - or are the red bars stacked in top of the blue bars?
- it's the latter
  - let's make is so that the blue histogram is layered on top of the red

```
ggplot(
  data = df_lexdec,
  aes(x = exp(RT), fill = NativeLanguage)
) +
  labs(title = "Stacked") +
  geom_histogram() +

ggplot(
  data = df_lexdec,
  aes(x = exp(RT), fill = NativeLanguage)
) +
  labs(title = "Layered: position = \"identity\"") +
  geom_histogram(position = "identity") +

plot_layout(guides = "collect") & theme(legend.position = 'bottom')
```



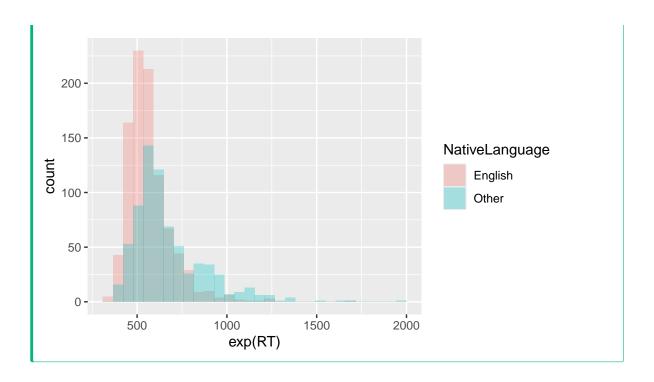
## 4.5 Global and local aesthetics

- in our final plot, the colour of the histograms is slightly transparent
  - we can control this by adding the argument alpha = 0.3 to geom\_histogram()
  - alpha takes any other value between 0 and 1

## • Aufgabe 4.2: Histogram transparency

#### Example 4.2.

Play around with the transparency of the histogram geom. Choose the alpha-value you prefer. The output should look something like this:



## 4.6 Customising our plot

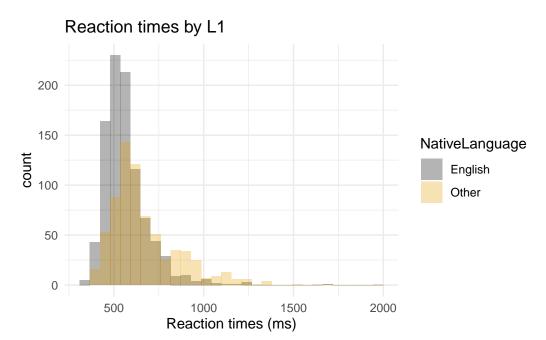
- we can improve our axis and legend labels, and also add titles using the labs() function
- let's also use the function scale\_fill\_colorblind() from the ggthemes package
  - this creates colourblind-safe colours
- we'll also use the theme\_minimal() function from ggplot2; what does this do?
- try to add the following to your plot
  - change the labels accordingly
  - and add meaningful comments to the code using #

```
labs(title = "Plot title",
    x = "x-axis label",
    y = "y-axis label") +
    scale_fill_colourblind() +
    theme_minimal()
```

## 4.7 Commenting

• the code and plot should look something like this:

```
# histogram of reaction times by native language
ggplot(data = df_lexdec) +
  aes(x = exp(RT), fill = NativeLanguage) + # set aesthetics
  labs(title = "Reaction times by L1",
       x = "Reaction times (ms)") +
  geom_histogram(position = "identity", alpha = 0.3) +
  scale_fill_colorblind() + # make fill colorblind friendly
  theme_minimal() # set plot theme
```



## 4.8 Saving plots

- we can store plots in our Environment, just like we can store numbers and data as objects
  - you can name objects anything you want
  - but it's wise to make the name meaningful (e.g., not fig1 or xyz)
- let's name this plot fig\_lexdec\_rt, for "figure lexical decision task reaction times"



• Aufgabe 4.3: ggplot2 review

## Example 4.3.

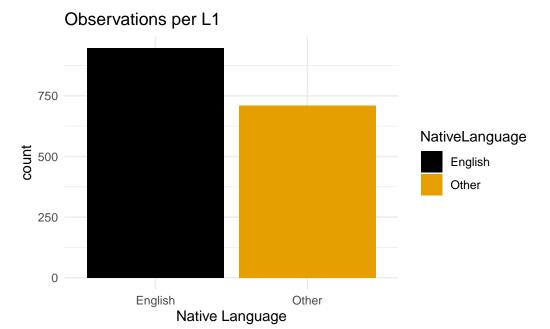
1. Save our final plot as an object called fig\_lexdec\_rt

## 4.9 Barplots

- copy the code for your histogram
- make the following changes to render our barplot
  - remove the name assignment (fig\_lexdec\_rt)
  - on the x-axis we want NativeLanguage
  - replace geom\_histogram() with geom\_bar()
    - \* remove the arguments for the histogram (not position or alpha)
  - change the labels accordingly
  - save the plot as an object with some meaningful name (e.g., fig\_lexdec\_l1)

## 4.10

• the plot should look something like this:

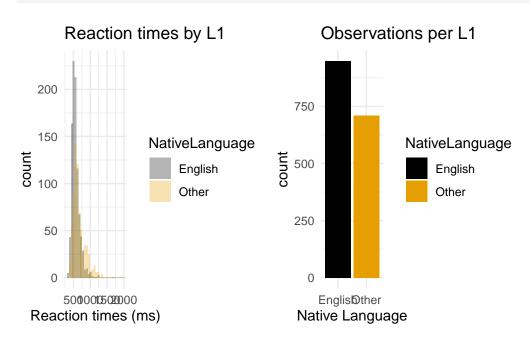


## 4.11 Combining plots

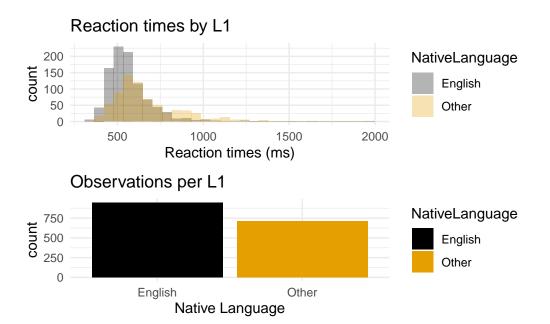
- one reason to save your plot as an object is so that we can call on it later
  - i.e., you can produce the plot at one point in your document, but decide to only print it in the rendered report lower down
- another reason is so that we can combine multiple plots
  - this can be done with a variety of packages
  - but let's try it with the patchwork package
  - use + to connect two plots side-by-side
  - or / to present them one on top of the other

#### 4.11.1 Combining plots with +

#### fig\_lexdec\_rt + fig\_lexdec\_l1



#### 4.11.2 Combining plots with /



## 5 Deciding on a geom

- why do we use histogram for reaction times, and a barplot for native language?
- what types of variables are these?
  - reaction time is continuous
  - native language is categorical
- we use histograms to visualise distributions of continuous variables
- we use barplots to visualise distributions of cateogrical variables
- knowing what we want to visualise (e.g., distributions) and what type of variable we have (i.e., continuous, categorical) helps us decide which type of plot to produce
- often, trying to draw your plot on paper before you start in R is a good idea (I often do this, too)

## 6 Exercises

These exercises should be also be included in your script if you upload it to Moodle. Working through the class materials will prepare you for these tasks.

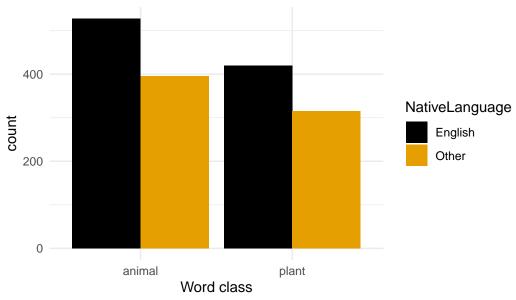
- 1. Reproduce our histogram as a *density plot* by replacing geom\_histogram() with geom\_density()
  - what does this type of plot show?

- 2. Produce a barplot that shows the number of observations per word class (hint: you'll need the variable Class from our dataset).
- 3. Print your density plot and class barplot one on top of the other using the patchwork package

## 6.1

4. Reproduce the following plots as exactly as you can (hint: you will need the position = "dodge" argument):

## Observations per word class/native language



## Heutige Ziele

Today we learned...

- what data frames are
- the difference between categorical and continuous data
- how to produce plots with ggplot
- choose the right plot for our data

#### Session Info

Hergestellt mit R version 4.4.0 (2024-04-24) (Puppy Cup) und RStudioversion 2023.3.0.386 (Cherry Blossom).

#### sessionInfo()

```
R version 4.4.0 (2024-04-24)
Platform: aarch64-apple-darwin20
Running under: macOS Ventura 13.2.1
Matrix products: default
        /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRblas.0.dylib
LAPACK: /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRlapack.dylib;
locale:
[1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
time zone: Europe/Berlin
tzcode source: internal
attached base packages:
[1] stats
              graphics
                        grDevices datasets utils
                                                       methods
                                                                 base
other attached packages:
 [1] magick_2.8.3
                      kableExtra_1.4.0 knitr_1.46
                                                         patchwork_1.2.0
                                        lubridate_1.9.3 forcats_1.0.0
 [5] ggthemes_5.1.0
                      languageR_1.5.0
 [9] stringr_1.5.1
                      dplyr_1.1.4
                                        purrr_1.0.2
                                                         readr_2.1.5
[13] tidyr_1.3.1
                      tibble_3.2.1
                                        ggplot2_3.5.1
                                                         tidyverse_2.0.0
loaded via a namespace (and not attached):
 [1] utf8_1.2.4
                       generics_0.1.3
                                          renv_1.0.7
                                                            xm12_1.3.6
 [5] stringi_1.8.3
                       hms_1.1.3
                                          digest_0.6.35
                                                            magrittr_2.0.3
 [9] evaluate_0.23
                       grid_4.4.0
                                          timechange_0.3.0
                                                            fastmap_1.1.1
[13] rprojroot_2.0.4
                                          tinytex_0.50
                       jsonlite_1.8.8
                                                            fansi_1.0.6
[17] viridisLite_0.4.2 scales_1.3.0
                                          cli_3.6.2
                                                            rlang_1.1.3
[21] munsell_0.5.1
                       withr_3.0.0
                                          yaml_2.3.8
                                                            tools_4.4.0
[25] tzdb_0.4.0
                       colorspace_2.1-0
                                         here_1.0.1
                                                            pacman_0.5.1
[29] vctrs_0.6.5
                       R6_2.5.1
                                          lifecycle_1.0.4
                                                            pkgconfig_2.0.3
[33] pillar_1.9.0
                       gtable_0.3.5
                                          Rcpp_1.0.12
                                                            glue_1.7.0
[37] systemfonts_1.0.6 xfun_0.43
                                                            rstudioapi_0.16.0
                                          tidyselect_1.2.1
[41] farver_2.1.1
                       htmltools_0.5.8.1 labeling_0.4.3
                                                            rmarkdown_2.26
```

[45] svglite\_2.1.3 compiler\_4.4.0

## Literaturverzeichnis

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