

# Data Transformation

Working with rows and columns

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## Inhaltsverzeichnis

<b>Wiederholung</b>	<b>2</b>
<b>Heutige Ziele</b>	<b>2</b>
<b>1 Pre-requisites</b>	<b>2</b>
<b>2 Data Wrangling</b>	<b>3</b>
2.1 lexdec . . . . .	4
2.2 dplyr basics . . . . .	4
<b>3 Rows</b>	<b>5</b>
3.1 filter() . . . . .	6
3.1.1 == and   . . . . .	7
3.1.2 %in% . . . . .	8
3.2 arrange() . . . . .	9
<b>4 Columns</b>	<b>11</b>
4.1 rename() . . . . .	11
4.2 mutate() . . . . .	12
4.3 Exercise . . . . .	13
4.4 select() . . . . .	13
4.5 select() helper functions . . . . .	15
4.6 relocate() . . . . .	16
<b>5 dplyr and ggplot2</b>	<b>17</b>
5.1 Exercises . . . . .	18
<b>Session Info</b>	<b>19</b>

# Wiederholung

Letzte Woche haben wir...

- gelernt, wie man einen neuen Datensatz in Augenschein nimmt
- gelernt, wie man verschiedene Datentypen importiert
- gelernt, wie man Daten von Hand eingibt
- einen neuen Datensatz visualisiert

# Heutige Ziele

Today we will...

- learn how to wrangle data using the `dplyr` package from the `tidyverse`
- learn to use the `pipe` (`|>`) to feed the result of one function into another function
- learn about functions that operate on rows
- learn about functions that operate on columns
- learn how to combine `dplyr` functions with plots from `ggplot2`

# Lust auf mehr?

- [Ch. 4](#) in Wickham et al. (o. J.)
- [Ch. 9](#) in Nordmann & DeBruine (2022)

# 1 Pre-requisites

## 1. Fresh Quarto document

- create a new Quarto document for today's class
  - File > New Document > Quarto Document, named something like 04-wrangling
- set up the YAML: title, your name, add a `toc`

```
title: "Data wrangling"
subtitle: "Transforming data"
author: "Your name here"
lang: de
date: "`r Sys.Date()`"
format:
  html:
```

```
toc: true
```

## 2. Packages

- today's packages are:
  - **tidyverse**: for wrangling (**dplyr**) and plotting (**ggplot2**)
  - **languageR**: for linguistic datasets

```
library(tidyverse)
library(languageR)
```

## 3. Data

- we're working again with the **lexdec** dataset from the **languageR** package (**languageR-package?**)
- store it as an object with the name **df\_lexdec**
- we also transform the **RT** variable so that it is in milliseconds (it was previously in log milliseconds, but don't worry about understanding what that means)
- and we choose 10 variables that are relevant for us today

```
df_lexdec <- lexdec |>
  mutate(RT = exp(RT)) |>
  select(Subject, RT, Trial, Sex, NativeLanguage, Correct, Word, Frequency, Class, Length)
```

# 2 Data Wrangling

- in English, wrangling refers to a long, difficult process
  - e.g., cowboys wrangle their cattle or herd (gather, collect their animals)
- there are two major parts of wrangling
  - transforming: sorting or creating new variables (what we'll do today)
  - tidying: reshaping or structuring your data (we'll do this in a few weeks)
- both data tidying and transforming require the **dplyr** package from the **tidyverse**
  - **dplyr** functions are often referred to as verbs, because they *do* something

## 2.1 lexdec

- the `lexdec` dataset contains data for a lexical decision task in English
  - let's take a look at the dataset using the `head()` function, which just prints the first 6 rows
    - \* here we're telling it to print the first 10 rows
- in my materials I often use the `head()` function to avoid printing the whole dataset in the output, but you wouldn't generally want to use `head()` when looking at your data, you'd want to look at your whole dataset

### Aufgabe 2.1: `df_lexdec`

#### Beispiel 2.1.

1. Look at the dataset
  - how many observations are there?
  - how many variables are there?
2. Feed the dataset into the `glimpse()` function
  - what does this show you?
  - how does it compare to what you see when you use `summary()`?

## 2.2 dplyr basics

- today we'll learn some of the primary `dplyr` verbs (functions) that allow us to solve the majority of our data manipulation challenges
  - I use these verbs multiple times in probably every analysis script
- `dplyr` verbs have some things in common:
  1. the first argument is always a data frame
  2. the following arguments typically describe which columns to be operated on, using the variable name (without quotation marks)
  3. the output is always a new dataframe
- the verbs all do one thing well, so we often want to use multiple verbs at once
  - we use the pipe to do this (`|>` or `|>`)
  - we've already seen this pipe when we feed a dataframe into `ggplot()`
  - we can read the pipe as **and then**

- in the following code, identify
  - the data frame
  - `dplyr` verbs
  - variable names
- can you try to read out (guess) what the following code does?

```
df_lexdec |>
  filter(Subject == "A1") |>
  select(Subject, Trial, RT, NativeLanguage, Word) |>
  relocate(NativeLanguage, .after = Trial)
```

- note that `A1` is written with quotation marks, but none of the other code is
  - when calling on an object (e.g., `df_lexdec`) or its variables (e.g., `Subject`), we do not wrap them in quotation marks
  - when we are calling on a certain *value* of a variable that is not numerical, we must wrap this value in quotation marks
  - because the Subject ID `A1` is a value of the variable `Subject`, we must use quotation marks around it
- try removing the quotation marks, what error message do you get?
- try adding quotation marks around a variable name, what error message do you get?
  - this is an important exercise, because you will often find your code will not run, but the solution is often something as simple as missing or extra quotation marks or punctuation

### 3 Rows

- in tidy data, rows represent observations
- the most important verbs for rows are:
  - `filter()`: changes which rows are present
  - `arrange()`: changes the order of rows
- we'll also discuss
  - `distinct()`: finds rows with distinct values based on a variable (column)

### 3.1 filter()

- changes which rows are present without changing their order
- takes the dataframe as first argument
  - following arguments are conditions that must be TRUE to keep the row
- find all reaction times that were longer than 450 milliseconds:

```
df_lexdec |>
  filter(RT > 450) |>
  head()
```

	Subject	RT	Trial	Sex	NativeLanguage	Correct	Word	Frequency	Class
1	A1	566.9998	23	F	English	correct	owl	4.859812	animal
2	A1	548.9998	27	F	English	correct	mole	4.605170	animal
3	A1	572.0000	29	F	English	correct	cherry	4.997212	plant
4	A1	486.0002	30	F	English	correct	pear	4.727388	plant
6	A1	483.0002	33	F	English	correct	blackberry	4.060443	plant
8	A1	524.9999	38	F	English	correct	squirrel	4.709530	animal
Length									
1	3								
2	4								
3	6								
4	4								
6	10								
8	8								

- notice that we don't put the reaction time value in quotation marks, because it is *numerical*
- if you want to save the filtered data, it's usually wise to save it with a *new* object name
  - unless you want to overwrite the pre-filtered version, a new name is necessary

```
df_lexdec_450 <-
  df_lexdec |>
  filter(RT > 450)
```

#### **i** Logical operators

- symbols used to describe a logical condition
- == *is identical* (1 == 1)

- `!=` is not identical (`1 != 2`)
- `>` is greater than (`2 > 1`)
- `<` is less than (`1 < 2`)
- to combine conditions
  - `&` or `,` and *also* (for multiple conditions)
  - `|` or (for multiple conditions)
- there's a nice shortcut for combining `==` and `|: %in%`
  - keeps rows where the variable equals one of the values on the right

### 3.1.1 `==` and `|`

```
df_lexdec |>
  filter(Trial == 30 | Trial == 23)
```

	Subject	RT	Trial	Sex	NativeLanguage	Correct	Word	Frequency
1	A1	566.9998	23	F	English	correct	owl	4.859812
4	A1	486.0002	30	F	English	correct	pear	4.727388
475	A2	561.0001	23	M	English	correct	dog	7.667626
949	C	688.0001	23	F	English	correct	vulture	4.248495
83	D	553.0000	30	M	Other	correct	walnut	4.499810
317	J	824.0004	23	F	Other	correct	beaver	3.951244
320	J	568.9998	30	F	Other	correct	carrot	4.976734
791	K	407.9999	23	F	English	correct	owl	4.859812
793	K	459.9998	30	F	English	correct	vulture	4.248495
1581	M2	941.9997	23	F	Other	incorrect	paprika	2.484907
1585	M2	628.9998	30	F	Other	correct	donkey	5.541264
159	P	1103.0000	23	F	Other	incorrect	moose	2.708050
1345	R1	483.0002	30	F	English	correct	ant	5.347108
1112	R2	601.0000	30	M	English	correct	snake	6.120297
1268	R3	422.9999	30	M	English	correct	dog	7.667626
558	T1	576.9998	30	F	English	correct	broccoli	2.833213
1423	V	1013.9998	23	F	Other	incorrect	stork	3.044522
241	Z	640.9997	30	M	Other	correct	squid	3.970292
Class Length								
1	animal	3						
4	plant	4						
475	animal	3						
949	animal	7						
83	plant	6						
317	animal	6						

```

320  plant      6
791  animal     3
793  animal     7
1581 plant      7
1585 animal     6
159  animal     5
1345 animal     3
1112 animal     5
1268 animal     3
558  plant      8
1423 animal     5
241  animal     5

```

### 3.1.2 %in%

```

df_lexdec |>
  filter(Trial %in% c(30, 23))

```

	Subject	RT	Trial	Sex	NativeLanguage	Correct	Word	Frequency
1	A1	566.9998	23	F	English	correct	owl	4.859812
4	A1	486.0002	30	F	English	correct	pear	4.727388
475	A2	561.0001	23	M	English	correct	dog	7.667626
949	C	688.0001	23	F	English	correct	vulture	4.248495
83	D	553.0000	30	M	Other	correct	walnut	4.499810
317	J	824.0004	23	F	Other	correct	beaver	3.951244
320	J	568.9998	30	F	Other	correct	carrot	4.976734
791	K	407.9999	23	F	English	correct	owl	4.859812
793	K	459.9998	30	F	English	correct	vulture	4.248495
1581	M2	941.9997	23	F	Other	incorrect	paprika	2.484907
1585	M2	628.9998	30	F	Other	correct	donkey	5.541264
159	P	1103.0000	23	F	Other	incorrect	moose	2.708050
1345	R1	483.0002	30	F	English	correct	ant	5.347108
1112	R2	601.0000	30	M	English	correct	snake	6.120297
1268	R3	422.9999	30	M	English	correct	dog	7.667626
558	T1	576.9998	30	F	English	correct	broccoli	2.833213
1423	V	1013.9998	23	F	Other	incorrect	stork	3.044522
241	Z	640.9997	30	M	Other	correct	squid	3.970292
Class Length								
1	animal	3						
4	plant	4						
475	animal	3						
949	animal	7						



```

83    plant      6
317   animal     6
320   plant      6
791   animal     3
793   animal     7
1581  plant      7
1585  animal     6
159   animal     5
1345  animal     3
1112  animal     5
1268  animal     3
558   plant      8
1423  animal     5
241   animal     5

```

### 💡 Aufgabe 3.1: filter()

#### Beispiel 3.1.

1. Filter the data to include rows from Trial 25 and non-native English speakers (other)
2. How many rows are there?

### 3.2 arrange()

- changes the order of the rows based on a value in a column(s)

```

df_lexdec |>
  arrange(RT) |>
  head()

```

	Subject	RT	Trial	Sex	NativeLanguage	Correct	Word	Frequency
542	A2	340.0001	159	M	English	incorrect	pig	6.660575
815	K	347.9998	83	F	English	incorrect	lemon	5.631212
822	K	363.0001	99	F	English	incorrect	potato	6.461468
73	A1	364.9999	174	F	English	correct	chicken	6.599870
524	A2	365.9999	117	M	English	correct	goose	5.267858
1516	I	367.0001	51	F	Other	correct	carrot	4.976734
	Class	Length						
542	animal	3						

```

815  plant      5
822  plant      6
73   animal     7
524  animal     5
1516 plant      6

```

- if you use more than one column name, each additional column will be used to break ties between values of the preceding columns

```

df_lexdec |>
  arrange(Length,Sex) |>
  head(10)

```

	Subject	RT	Trial	Sex	NativeLanguage	Correct	Word	Frequency	Class
1	A1	566.9998	23	F	English	correct	owl	4.859812	animal
5	A1	414.0000	32	F	English	correct	dog	7.667626	animal
15	A1	556.9999	53	F	English	correct	bee	5.700444	animal
20	A1	456.9998	61	F	English	incorrect	bat	5.918894	animal
31	A1	581.9997	88	F	English	correct	fox	5.652489	animal
44	A1	494.0002	113	F	English	correct	pig	6.660575	animal
62	A1	467.9999	152	F	English	correct	cat	7.086738	animal
64	A1	875.9999	157	F	English	correct	ant	5.347108	animal
719	A3	607.0001	41	F	Other	correct	ant	5.347108	animal
720	A3	562.0001	44	F	Other	correct	pig	6.660575	animal
Length									
1	3								
5	3								
15	3								
20	3								
31	3								
44	3								
62	3								
64	3								
719	3								
720	3								

- we can add `desc()` inside `arrange()` to use descending order (big-to-small) instead of the default ascending order

```

df_lexdec |>
  arrange(desc(Length)) |>
  head()

```

	Subject	RT	Trial	Sex	NativeLanguage	Correct	Word	Frequency
6	A1	483.0002	33	F	English	correct	blackberry	4.060443
7	A1	417.9998	34	F	English	correct	strawberry	4.753590
69	A1	540.9998	168	F	English	correct	woodpecker	2.890372
505	A2	503.9999	87	M	English	correct	woodpecker	2.890372
516	A2	400.9998	105	M	English	correct	strawberry	4.753590
518	A2	517.0001	108	M	English	correct	blackberry	4.060443
	Class	Length						
6	plant	10						
7	plant	10						
69	animal	10						
505	animal	10						
516	plant	10						
518	plant	10						

### 💡 Aufgabe 3.2: `arrange()`

#### Beispiel 3.2.

1. Filter the data to include observations from only the **Subject** M1 and W2, *and then*
2. Arrange the data by descending reaction times

## 4 Columns

- in tidy data, columns represent variables
- the most important verbs for columns are:
  - `rename()`: changes the names of the columns
  - `mutate()`: creates new columns that are derived from the existing columns
  - `select()`: changes which columns are present
  - `relocate()`: changes the positions of the columns

### 4.1 `rename()`

- `rename()` lets us change the name of columns
  - the order of the arguments is `new_name = old_name`
- let's try changing some of the variable names to German

- I tend to create variable names with lower case, as a coding convention

```
# single variable
df_lexent <-
  df_lexdec |>
  rename(teilnehmer = Subject)

# or multiple variables at once
df_lexent <-
  df_lexdec |>
  rename(teilnehmer = Subject,
         rz_ms = RT,
         geschlect = Sex,
         laenge = Length)
```

## 4.2 mutate()

- `mutate()` creates new columns from existing columns
  - e.g., we can perform basic algebra on the values in each column

```
df_lexent |>
  mutate(
    rz_laenge = rz_ms / laenge,
  ) |>
  head()
```

	teilnehmer	rz_ms	Trial	geschlect	NativeLanguage	Correct	Word
1	A1	566.9998	23	F	English	correct	owl
2	A1	548.9998	27	F	English	correct	mole
3	A1	572.0000	29	F	English	correct	cherry
4	A1	486.0002	30	F	English	correct	pear
5	A1	414.0000	32	F	English	correct	dog
6	A1	483.0002	33	F	English	correct	blackberry

	Frequency	Class	laenge	rz_laenge
1	4.859812	animal	3	188.99994
2	4.605170	animal	4	137.24994
3	4.997212	plant	6	95.33333
4	4.727388	plant	4	121.50005
5	7.667626	animal	3	138.00000
6	4.060443	plant	10	48.30002

- `mutate()` adds these new columns to the right of your dataset
  - this makes it difficult to see what's happening
- to control where the new column is added, we can use `.before` or `.after`

```
df_lexent |>
  mutate(
    rz_laenge = rz_ms / laenge,
    .after = rz_ms
  ) |>
  head()
```

	teilnehmer	rz_ms	rz_laenge	Trial	geschlect	NativeLanguage	Correct
1	A1	566.9998	188.99994	23	F	English	correct
2	A1	548.9998	137.24994	27	F	English	correct
3	A1	572.0000	95.33333	29	F	English	correct
4	A1	486.0002	121.50005	30	F	English	correct
5	A1	414.0000	138.00000	32	F	English	correct
6	A1	483.0002	48.30002	33	F	English	correct

	Word	Frequency	Class	laenge
1	owl	4.859812	animal	3
2	mole	4.605170	animal	4
3	cherry	4.997212	plant	6
4	pear	4.727388	plant	4
5	dog	7.667626	animal	3
6	blackberry	4.060443	plant	10

### 4.3 Exercise

1. Create a new variable called `rz_s` in `df_lexent`:
  - equals `rz_ms` divided by 1000 (i.e., converts milliseconds to seconds)
  - appears after `rz_ms`
2. Render your document

### 4.4 `select()`

- `select()` subsets the data to include only the columns you want
- select columns by name

```
df_lexent |>
  select(teilnehmer, rz_ms, Word) |>
  head()
```

	teilnehmer	rz_ms	Word
1	A1	566.9998	owl
2	A1	548.9998	mole
3	A1	572.0000	cherry
4	A1	486.0002	pear
5	A1	414.0000	dog
6	A1	483.0002	blackberry

- select all columns between rz\_ms and geschlecht

```
df_lexent |>
  select(rz_ms:geschlecht) |>
  head()
```

	rz_ms	rz_s	Trial	geschlecht
1	566.9998	0.5669998	23	F
2	548.9998	0.5489998	27	F
3	572.0000	0.5720000	29	F
4	486.0002	0.4860002	30	F
5	414.0000	0.4140000	32	F
6	483.0002	0.4830002	33	F

- select all columns except rz\_s (! is read as “not”)

```
df_lexent |>
  select(!rz_s) |>
  head()
```

	teilnehmer	rz_ms	Trial	geschlecht	NativeLanguage	Correct	Word
1	A1	566.9998	23	F	English	correct	owl
2	A1	548.9998	27	F	English	correct	mole
3	A1	572.0000	29	F	English	correct	cherry
4	A1	486.0002	30	F	English	correct	pear
5	A1	414.0000	32	F	English	correct	dog
6	A1	483.0002	33	F	English	correct	blackberry

Frequency Class laenge

1	4.859812	animal	3
2	4.605170	animal	4
3	4.997212	plant	6
4	4.727388	plant	4
5	7.667626	animal	3
6	4.060443	plant	10

## 4.5 select() helper functions

- some helper functions that make life easier when working with `select()`:
  - `starts_with("abc")`: selects columns that begin with a certain string of characters
  - `ends_with("xyz")`: selects columns that end with a certain string of characters
  - `contains("ijk")`: selects columns that contain a certain string of characters
  - `where(is.character)`: selects columns that match a logical criteria
    - \* e.g., the function `is.character()` returns the value `TRUE` when a variable contains character strings, not numerical values or categories

```
df_lexent |>
  select(starts_with("w")) |>
  head()
```

	Word
1	owl
2	mole
3	cherry
4	pear
5	dog
6	blackberry

```
df_lexent |>
  select(ends_with("er")) |>
  head()
```

	teilnehmer
1	A1
2	A1
3	A1
4	A1
5	A1
6	A1

#### 💡 Aufgabe 4.1: `select()`

##### Beispiel 4.1.

1. Print the columns in `df_lexent` that begin with “t”
2. Print the columns in `df_lexent` that contain “ge”
3. Print the columns in `df_lexent` that
  - begin with begin with “r”, and
  - end with “s”

## 4.6 `relocate()`

- `relocate()` moves variables around
  - by default, it moves them to the front

```
df_lexent |> relocate(Trial) |>
  head()
```

	Trial	teilnehmer		rz_ms	rz_s	geschlect	NativeLanguage	Correct
1	23	A1	566.9998	0.5669998		F	English	correct
2	27	A1	548.9998	0.5489998		F	English	correct
3	29	A1	572.0000	0.5720000		F	English	correct
4	30	A1	486.0002	0.4860002		F	English	correct
5	32	A1	414.0000	0.4140000		F	English	correct
6	33	A1	483.0002	0.4830002		F	English	correct
		Word	Frequency	Class	laenge			
1		owl	4.859812	animal	3			
2		mole	4.605170	animal	4			
3		cherry	4.997212	plant	6			
4		pear	4.727388	plant	4			
5		dog	7.667626	animal	3			
6		blackberry	4.060443	plant	10			

- but we can also use `.before` or `.after` to place a variable

```
df_lexent |>
  relocate(Trial, .after = teilnehmer) |>
  head()
```



	teilnehmer	Trial	rz_ms	rz_s	geschlect	NativeLanguage	Correct
1	A1	23	566.9998	0.5669998	F	English	correct
2	A1	27	548.9998	0.5489998	F	English	correct
3	A1	29	572.0000	0.5720000	F	English	correct
4	A1	30	486.0002	0.4860002	F	English	correct
5	A1	32	414.0000	0.4140000	F	English	correct
6	A1	33	483.0002	0.4830002	F	English	correct

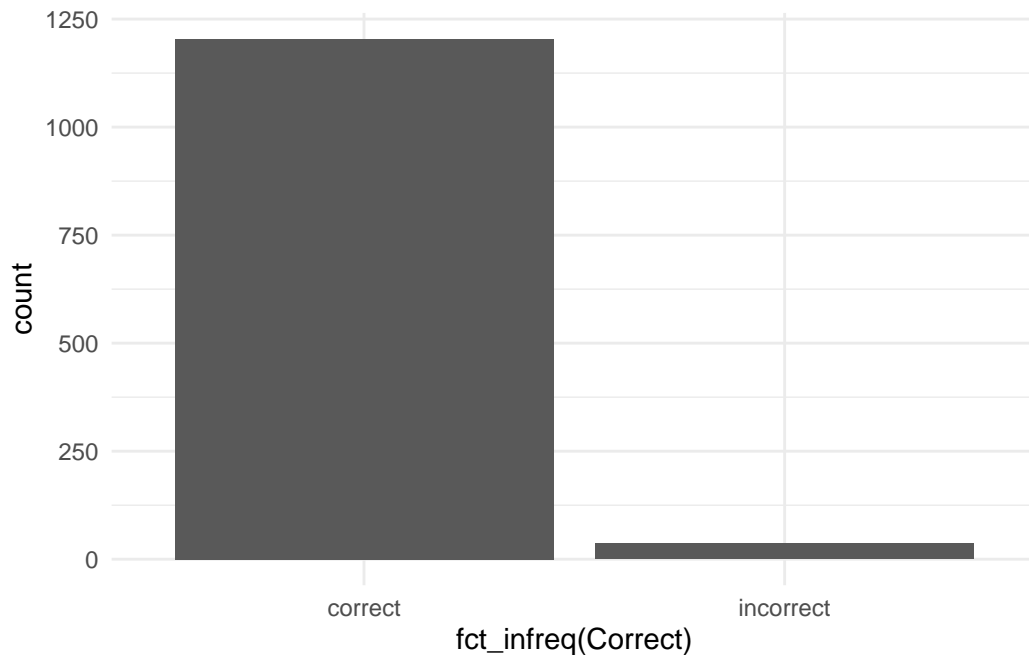
  

	Word	Frequency	Class	laenge
1	owl	4.859812	animal	3
2	mole	4.605170	animal	4
3	cherry	4.997212	plant	6
4	pear	4.727388	plant	4
5	dog	7.667626	animal	3
6	blackberry	4.060443	plant	10

## 5 dplyr and ggplot2

- we can change a dataset using the `dplyr` verbs, and then feed these changes into `ggplot2`
- what will the following code produce?

```
df_lexent |>
  # filter the data
  filter(rz_ms > 120,
         rz_ms > 500) |>
  # plot the filtered data
  ggplot(aes(x = fct_infreq(Correct))) +
  geom_bar() +
  theme_minimal()
```



- important: we can use pipes (`|>`) to perform additional verbs/functions
  - but the `ggplot()` function uses `+` to add new *layers* to the plot

## 5.1 Exercises

1. In a single pipeline, print `df_lexent` where you **select** only the reaction times (in milliseconds), `NativeLanguage`, and `Word` columns for rows that meet each of the following conditions, **arrange** them in order of reaction times, and **filter** them to include only:
  - reaction times were greater than 500ms *and* less than 550ms
  - were from the words “pear”, “elephant”, or “tortoise”
2. Sort `df_lexent` in descending order to find the trials with longest reaction times.
3. In a single pipeline, store a new object called `df_rz` which contains `df_lexent`, *and then*:
  - select the variables `teilnehmer`, `NativeLanguage`, `Word`, `rz_s`, `laenge`, and `Frequency`
  - create a new variable `rz_s_laenge`, that is `rz_s` divided by `laenge`
    - and is placed before `laenge`
  - rename these variables in English so that they are in German (and with lower case)

## Heutige Ziele

Today we learned...

- learn how to wrangle data using the `dplyr` package from the `tidyverse`
- learn to use the `pipe` (`|>`) to feed the result of one function into another function
- learn about functions that operate on rows
- learn about functions that operate on columns
- learn how to combine `dplyr` functions with plots from `ggplot2`

## Session Info

Hergestellt mit R version 4.3.0 (2023-04-21) (Already Tomorrow) und RStudioversion 2023.3.0.386 (Cherry Blossom).

```
sessionInfo()
```

R version 4.3.0 (2023-04-21)

Platform: aarch64-apple-darwin20 (64-bit)

Running under: macOS Ventura 13.2.1

Matrix products: default

BLAS: /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/lib/libRblas.0.dylib

LAPACK: /Library/Frameworks/R.framework/Versions/4.3-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 utils datasets methods base

other attached packages:

[1] languageR\_1.5.0 lubridate\_1.9.2 forcats\_1.0.0 stringr\_1.5.0

[5] dplyr\_1.1.3 purrr\_1.0.2 readr\_2.1.4 tidyr\_1.3.0

[9] tibble\_3.2.1 ggplot2\_3.4.3 tidyverse\_2.0.0

loaded via a namespace (and not attached):

[1] gtable\_0.3.4 jsonlite\_1.8.7 compiler\_4.3.0 tidyselect\_1.2.0

[5]	scales_1.2.1	yaml_2.3.7	fastmap_1.1.1	R6_2.5.1
[9]	labeling_0.4.3	generics_0.1.3	knitr_1.44	munSELL_0.5.0
[13]	pillar_1.9.0	tzdb_0.4.0	rlang_1.1.1	utf8_1.2.3
[17]	stringi_1.7.12	xfun_0.39	timechange_0.2.0	cli_3.6.1
[21]	withr_2.5.0	magrittr_2.0.3	digest_0.6.33	grid_4.3.0
[25]	rstudioapi_0.14	hms_1.1.3	lifecycle_1.0.3	vcTrs_0.6.3
[29]	evaluate_0.21	glue_1.6.2	farver_2.1.1	fansi_1.0.4
[33]	colorspace_2.1-0	rmarkdown_2.22	tools_4.3.0	pkgconfig_2.0.3
[37]	htmltools_0.5.5			

Nordmann, E., & DeBruine, L. (2022). *Applied Data Skills* (Version 2.0). Zenodo. <https://doi.org/10.5281/zenodo.6365078>

Wickham, H., Çetinkaya-Rundel, M., & Grolemund, G. (o. J.). *R for Data Science* (2. Aufl.). <https://r4ds.hadley.nz/>