

Working with eye-tracking reading data in R

Loading and eye-balling a dataset

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Set-up

```
knitr::opts_chunk$set(eval = T, # evaluate = T for REPRODUCIBLE analyses
  echo = T, # 'print code chunk?'
  message = F, # print messages?
  error = T, # render even if errors encountered?
  warning = F) # print warnings?

1 library(here) # relative path
2 library(tidyverse) # tidy/transform
3 library(beep) # beeps when code runs or fails
4 library(rbbt) # zotero plugin

## play sound if error encountered
### from: https://sejohnston.com/2015/02/24/make-r-beep-when-r-markdown-finishes-or-when-i
options(error = function(){ # Beep on error
  beepr::beep(sound = "wilhelm")
  Sys.sleep(2) #
})
## and when knitting is complete
.Last <- function() { # Beep on exiting session
  beepr::beep(sound = "ping")
  Sys.sleep(6) # allow to play for 6 seconds
}

# Create references.json file based on the citations in this script:
# 1. make sure you have 'bibliography: references/references.json' in the YAML
# 2. create a new folder called 'references'
# 3. run:
rbbt::bibt_update_bib("_et_dataset.qmd")
```

The Perfect Lifetime Effect

- the English Present Perfect (e.g., *has done*) (e.g., Comrie, 1976)
 - must be used in temporal contexts that *include the present*
 - * *I have been sick since last week*
 - * **I have been sick last year*

- The Lifetime Effect
 - a referent’s lifetime (dead/alive) constrains verb tense in certain circumstances (e.g., Mittwoch, 2008a)
 - * **Queen Elizabeth II is the British monarch.*
 - * **King Charles III was the British monarch.*
- the Perfect Lifetime Effect
 - the (English) Present Perfect cannot be used to describe events of a dead person (e.g., Mittwoch, 2008b)
 - * **Queen Elizabeth II has met many politicians.*
 - * **King Charles III has met many politicians.*

Our first dataset

- referent-lifetime context
 - dead/alive
- critical sentence
 - Present Perfect/Simple Future
- binary naturalness judgement to end trial
 - accept/reject

Design description

- 2x2 mixed design
 - two 2-level factors (2x2 = 2-level x 2-level)
 - * factor 1: lifetime (levels: dead, alive)
 - * factor 2: tense (levels: PP, SF)

	alive	dead
PP	Eddie Redmayne...has won	Gene Kelly...*has won
SF	Eddie Redmayne...will win	Gene Kelly...*will win

- predictors/independent variables
 - lifetime
 - tense

- measure/dependent variables (verb region)
 - first-fixation time (milliseconds)
 - first-pass reading time (ms)
 - regression path duration (ms)
 - total reading time (ms)

Repeated measures design

- observations are repeated e.g., multiple data points per participant, and per item across participants
 - essentially, data are not independent
 - e.g., each participant will have their own reading speed, some items might be systematically less acceptable for some unforeseen reason, etc.

Working with the data

Day 1

1. load the data
2. inspect data
 - eyeball data structure
 - print summaries
 - plot data distributions

Day 2

3. tidy data
4. visualise data
5. communicate data

Day 3

6. analyse data
 - confirmatory (a priori)
 - exploratory (post-hoc)
7. report analyses

Install packages

```
install.packages("tidyverse")  
install.packages("here")
```

- install
 - only do once
 - ...or when you working on a new computer
 - ...or after updating R
- might be a wise idea to create a script just for installing packages
 - can save time/energy when updating R

Load packages

```
library(tidyverse)  
library(here)
```

- load packages
 - needed at the start of each session

Load dataset

```
df_lifetime <- readr::read_csv(here::here("data/data_lifetime_pilot.csv"))
```

- N.B., `readr::read_csv` can be read as “`read_csv()` function in the `readr` package”
 - i.e., `package::function()`
 - you only need to use this syntax if you haven’t loaded the specific package yet (maybe because you only need it once), or if a function name is included in multiple packages (i.e., there’s a discrepancy in what `read_csv` could be referring to)
 - why did I use it here?

💡 **here** package

Using the **here** package, we can access files *relative* to where our .RProj is stored. In ‘olden times’, we had to specify the file path with something like:

```
# load in data from an *absolute* file path
df_lifetime <- read_csv("Users/yournamehere/Documents/SoSe2023/ET_reading/data/data_lifetime.csv")
```

Or, we’d set an *absolute* path as our working directory, to which all other file paths were *relative*

```
# set *absolute* path as working directory
setwd("Users/username/Documents/SoSe2023/ET_reading")

# load in data *relative* to our wd
df_lifetime <- read_csv("data/data_lifetime_pilot.csv")
```

This meant that if I sent my project folder to somebody else, they wouldn’t be able to run my code because they would have to change the *absolute* file path to match their machine.

Inspect dataset

- there are several different things you can inspect
 - and different ways to accomplish those things
- the first thing I usually do is look at the column/variable names

names()

- the names in all caps are variables created during the experiment
 - i.e., they are our recorded *data*, mainly what we wanted to measure: dependent variables (DV)
 - also includes some information about the experiment set-up per participant
- the other names are variables from my stimuli lists
 - i.e., they mostly contain our independent variables (IV)/stimuli
- we typically want to see what effect our IVs had on any given DVs

- variable descriptions can be found on the Moodle: Data > Documentation

```
names(df_lifetime)
```

```
[1] "RECORDING_SESSION_LABEL"      "TRIAL_INDEX"
[3] "EYE_USED"                     "IA_DWELL_TIME"
[5] "IA_FIRST_FIXATION_DURATION"   "IA_FIRST_RUN_DWELL_TIME"
[7] "IA_FIXATION_COUNT"           "IA_ID"
[9] "IA_LABEL"                     "IA_REGRESSION_IN"
[11] "IA_REGRESSION_IN_COUNT"       "IA_REGRESSION_OUT"
[13] "IA_REGRESSION_OUT_COUNT"      "IA_REGRESSION_PATH_DURATION"
[15] "KeyPress"                     "rt"
[17] "bio"                           "critical"
[19] "gender"                       "item_id"
[21] "list"                         "match"
[23] "condition"                    "name"
[25] "name_vital_status"           "tense"
[27] "type"                         "yes_press"
```

```
rename()
```

- the dependent variable names are pretty clunky, let's rename a few:
 - RECORDING_SESSION_LABEL corresponds to a single participant
 - TRIAL_INDEX logged the trial number
 - EYE_USED logged which eye was tracked

```
df_lifetime <- df_lifetime %>%
  rename("px" = RECORDING_SESSION_LABEL,
         "trial" = TRIAL_INDEX,
         "eye" = EYE_USED)
```

Naming variables

Naming conventions

It's wise to keep variable and object names concise but informative

- all lowercase means fewer key strokes overall
- separate words with either periods or underscores, e.g., `trial.index` or `trial_index`

- e.g., we called our dataset `df_lifetime` because it is a dataframe (`df`) with data from our lifetime experiment

Data structure

- datasets typically contain a lot of rows and columns
 - so we want to get a feel for how the data is structured

with base R

```
head(df_lifetime)
```

```
# A tibble: 6 x 28
  px    trial eye IA_DWELL_TIME IA_FIRST_FIXATION_DUR~1 IA_FIRST_RUN_DWELL_T~2
  <chr> <dbl> <chr>      <dbl>                <dbl>                <dbl>
1 px3      1 RIGHT          0                    0                    0
2 px3      2 RIGHT          0                    0                    0
3 px3      3 RIGHT          0                    0                    0
4 px3      3 RIGHT          0                    0                    0
5 px3      3 RIGHT          0                    0                    0
6 px3      3 RIGHT          0                    0                    0
# i abbreviated names: 1: IA_FIRST_FIXATION_DURATION,
#   2: IA_FIRST_RUN_DWELL_TIME
# i 22 more variables: IA_FIXATION_COUNT <dbl>, IA_ID <dbl>, IA_LABEL <chr>,
#   IA_REGRESSION_IN <dbl>, IA_REGRESSION_IN_COUNT <dbl>,
#   IA_REGRESSION_OUT <dbl>, IA_REGRESSION_OUT_COUNT <dbl>,
#   IA_REGRESSION_PATH_DURATION <dbl>, KeyPress <dbl>, rt <dbl>, bio <chr>,
#   critical <chr>, gender <chr>, item_id <dbl>, list <dbl>, match <chr>, ...
```

with the tidyverse pipe

```
1 df_lifetime %>%
2   head()
```

```
# A tibble: 6 x 28
  px    trial eye IA_DWELL_TIME IA_FIRST_FIXATION_DUR~1 IA_FIRST_RUN_DWELL_T~2
  <chr> <dbl> <chr>      <dbl>                <dbl>                <dbl>
1 px3      1 RIGHT          0                    0                    0
```



```

2 px3      2 RIGHT      0      0      0
3 px3      3 RIGHT      0      0      0
4 px3      3 RIGHT      0      0      0
5 px3      3 RIGHT      0      0      0
6 px3      3 RIGHT      0      0      0
# i abbreviated names: 1: IA_FIRST_FIXATION_DURATION,
#   2: IA_FIRST_RUN_DWELL_TIME
# i 22 more variables: IA_FIXATION_COUNT <dbl>, IA_ID <dbl>, IA_LABEL <chr>,
#   IA_REGRESSION_IN <dbl>, IA_REGRESSION_IN_COUNT <dbl>,
#   IA_REGRESSION_OUT <dbl>, IA_REGRESSION_OUT_COUNT <dbl>,
#   IA_REGRESSION_PATH_DURATION <dbl>, KeyPress <dbl>, rt <dbl>, bio <chr>,
#   critical <chr>, gender <chr>, item_id <dbl>, list <dbl>, match <chr>, ...

```

with the native R pipe (Ctrl/Cmd+Shift+M)

```

1 df_lifetime |>
2   head()

# A tibble: 6 x 28
  px    trial eye  IA_DWELL_TIME IA_FIRST_FIXATION_DUR~1 IA_FIRST_RUN_DWELL_T~2
  <chr> <dbl> <chr>      <dbl>              <dbl>              <dbl>
1 px3      1 RIGHT      0                0                0
2 px3      2 RIGHT      0                0                0
3 px3      3 RIGHT      0                0                0
4 px3      3 RIGHT      0                0                0
5 px3      3 RIGHT      0                0                0
6 px3      3 RIGHT      0                0                0
# i abbreviated names: 1: IA_FIRST_FIXATION_DURATION,
#   2: IA_FIRST_RUN_DWELL_TIME
# i 22 more variables: IA_FIXATION_COUNT <dbl>, IA_ID <dbl>, IA_LABEL <chr>,
#   IA_REGRESSION_IN <dbl>, IA_REGRESSION_IN_COUNT <dbl>,
#   IA_REGRESSION_OUT <dbl>, IA_REGRESSION_OUT_COUNT <dbl>,
#   IA_REGRESSION_PATH_DURATION <dbl>, KeyPress <dbl>, rt <dbl>, bio <chr>,
#   critical <chr>, gender <chr>, item_id <dbl>, list <dbl>, match <chr>, ...

```

head() function

- *prints* the first 6 rows of your data
 - you can also specify the number of rows

```
df_lifetime %>%
  head(n = 2)
```

```
# A tibble: 2 x 28
  px    trial eye  IA_DWELL_TIME IA_FIRST_FIXATION_DUR~1 IA_FIRST_RUN_DWELL_T~2
  <chr> <dbl> <chr>      <dbl>              <dbl>              <dbl>
1 px3      1 RIGHT          0                  0                  0
2 px3      2 RIGHT          0                  0                  0
# i abbreviated names: 1: IA_FIRST_FIXATION_DURATION,
#   2: IA_FIRST_RUN_DWELL_TIME
# i 22 more variables: IA_FIXATION_COUNT <dbl>, IA_ID <dbl>, IA_LABEL <chr>,
#   IA_REGRESSION_IN <dbl>, IA_REGRESSION_IN_COUNT <dbl>,
#   IA_REGRESSION_OUT <dbl>, IA_REGRESSION_OUT_COUNT <dbl>,
#   IA_REGRESSION_PATH_DURATION <dbl>, KeyPress <dbl>, rt <dbl>, bio <chr>,
#   critical <chr>, gender <chr>, item_id <dbl>, list <dbl>, match <chr>, ...
```

head() function task

Exercise: head()

1. print only 2 rows using whichever syntax you prefer
2. change `n = 2` to some other number and print
3. run `?head` in the Console
 - find the opposite function (i.e., prints last rows) in the function description?
4. run this function with `df_lifetime` as argument; how many rows does it print as default?
5. play with `n =` in this function to print some other number of rows

tail() function

- prints the last rows of a dataframe (or matrix, vector, table, or function)

```
df_lifetime %>%
  tail()
```

```
# A tibble: 6 x 28
  px    trial eye  IA_DWELL_TIME IA_FIRST_FIXATION_DUR~1 IA_FIRST_RUN_DWELL_T~2
  <chr> <dbl> <chr>      <dbl>              <dbl>              <dbl>
```

```

1 px4      207 LEFT      509      218      509
2 px4      208 LEFT      0        0        0
3 px4      208 LEFT      317      167      317
4 px4      208 LEFT      162      162      162
5 px4      208 LEFT      139      139      139
6 px4      208 LEFT      280      280      280
# i abbreviated names: 1: IA_FIRST_FIXATION_DURATION,
#   2: IA_FIRST_RUN_DWELL_TIME
# i 22 more variables: IA_FIXATION_COUNT <dbl>, IA_ID <dbl>, IA_LABEL <chr>,
#   IA_REGRESSION_IN <dbl>, IA_REGRESSION_IN_COUNT <dbl>,
#   IA_REGRESSION_OUT <dbl>, IA_REGRESSION_OUT_COUNT <dbl>,
#   IA_REGRESSION_PATH_DURATION <dbl>, KeyPress <dbl>, rt <dbl>, bio <chr>,
#   critical <chr>, gender <chr>, item_id <dbl>, list <dbl>, match <chr>, ...

```

`names()`

- prints the column/variable names

```

df_lifetime %>%
  names()

```

```

[1] "px"                "trial"
[3] "eye"               "IA_DWELL_TIME"
[5] "IA_FIRST_FIXATION_DURATION" "IA_FIRST_RUN_DWELL_TIME"
[7] "IA_FIXATION_COUNT"    "IA_ID"
[9] "IA_LABEL"            "IA_REGRESSION_IN"
[11] "IA_REGRESSION_IN_COUNT" "IA_REGRESSION_OUT"
[13] "IA_REGRESSION_OUT_COUNT" "IA_REGRESSION_PATH_DURATION"
[15] "KeyPress"           "rt"
[17] "bio"                "critical"
[19] "gender"             "item_id"
[21] "list"               "match"
[23] "condition"          "name"
[25] "name_vital_status"  "tense"
[27] "type"               "yes_press"

```

`summary()`

- prints a summary of each variable (column)

```
df_lifetime %>%
summary()
```

px	trial	eye	IA_DWELL_TIME
Length:4431	Min. : 1.0	Length:4431	Min. : 0.0
Class :character	1st Qu.: 52.5	Class :character	1st Qu.: 0.0
Mode :character	Median :104.0	Mode :character	Median : 301.0
	Mean :105.0		Mean : 587.5
	3rd Qu.:157.0		3rd Qu.: 765.5
	Max. :208.0		Max. :8968.0
IA_FIRST_FIXATION_DURATION	IA_FIRST_RUN_DWELL_TIME	IA_FIXATION_COUNT	
Min. : 0.0	Min. : 0.0	Min. : 0.000	
1st Qu.: 0.0	1st Qu.: 0.0	1st Qu.: 0.000	
Median :161.0	Median : 245.0	Median : 2.000	
Mean :139.4	Mean : 507.9	Mean : 2.714	
3rd Qu.:202.5	3rd Qu.: 586.0	3rd Qu.: 4.000	
Max. :775.0	Max. :8968.0	Max. :35.000	
IA_ID	IA_LABEL	IA_REGRESSION_IN	IA_REGRESSION_IN_COUNT
Min. :1.000	Length:4431	Min. :0.00000	Min. :0.0000
1st Qu.:1.000	Class :character	1st Qu.:0.00000	1st Qu.:0.0000
Median :2.000	Mode :character	Median :0.00000	Median :0.0000
Mean :2.681		Mean :0.09817	Mean :0.1318
3rd Qu.:4.000		3rd Qu.:0.00000	3rd Qu.:0.0000
Max. :6.000		Max. :1.00000	Max. :5.0000
IA_REGRESSION_OUT	IA_REGRESSION_OUT_COUNT	IA_REGRESSION_PATH_DURATION	
Min. :0.00000	Min. :0.00000	Min. : 0.0	
1st Qu.:0.00000	1st Qu.:0.00000	1st Qu.: 0.0	
Median :0.00000	Median :0.00000	Median : 282.0	
Mean :0.08147	Mean :0.09185	Mean : 595.6	
3rd Qu.:0.00000	3rd Qu.:0.00000	3rd Qu.: 747.0	
Max. :1.00000	Max. :7.00000	Max. :10242.0	
KeyPress	rt	bio	critical
Min. :4.000	Min. : 533	Length:4431	Length:4431
1st Qu.:4.000	1st Qu.: 1332	Class :character	Class :character
Median :4.000	Median : 1890	Mode :character	Mode :character
Mean :4.496	Mean : 2467		
3rd Qu.:5.000	3rd Qu.: 2910		
Max. :5.000	Max. :15654		
gender	item_id	list	match
Length:4431	Min. : 1.00	Min. :14.00	Length:4431
Class :character	1st Qu.: 26.00	1st Qu.:15.00	Class :character
Mode :character	Median : 51.00	Median :25.00	Mode :character

	Mean : 64.16	Mean :29.45	
	3rd Qu.: 78.50	3rd Qu.:35.00	
	Max. :208.00	Max. :45.00	
condition	name	name_vital_status	tense
Length:4431	Length:4431	Length:4431	Length:4431
Class :character	Class :character	Class :character	Class :character
Mode :character	Mode :character	Mode :character	Mode :character

type	yes_press
Length:4431	Min. :4.000
Class :character	1st Qu.:4.000
Mode :character	Median :4.000
	Mean :4.499
	3rd Qu.:5.000
	Max. :5.000

Exercise

Take some time to explore the dataset.

- double click on the dataset name in the Environment pane to view it like a spreadsheet
- look at the names, can you figure out what they represent?

class types

- there are difference classes of data that R can read
 - the function `class()` takes as its argument an object or number

```
1 df_lifetime$rt %>%
2   class()
```

```
[1] "numeric"
```

💡 Selecting a column

```
# with column index  
df_lifetime[2] %>% summary()
```

```
trial  
Min.   : 1.0  
1st Qu.: 52.5  
Median :104.0  
Mean   :105.0  
3rd Qu.:157.0  
Max.   :208.0
```

```
# with column name  
df_lifetime[, "trial"] %>% summary()
```

```
trial  
Min.   : 1.0  
1st Qu.: 52.5  
Median :104.0  
Mean   :105.0  
3rd Qu.:157.0  
Max.   :208.0
```

```
# with data$column_name  
df_lifetime$trial %>% summary()
```

```
Min. 1st Qu.  Median    Mean 3rd Qu.    Max.  
1.0   52.5   104.0   105.0  157.0   208.0
```

```
# with the tidyverse: select()  
df_lifetime %>%  
  select(trial) %>%  
  summary()
```

```
trial  
Min.   : 1.0  
1st Qu.: 52.5
```

```
Median :104.0
Mean   :105.0
3rd Qu.:157.0
Max.    :208.0
```

character class

- contain *strings*: collection of characters (i.e., text)
- there's no grouping in character variables
 - each value is considered 'unique' and assumed to not be repeated
- we usually aren't interested in character class variables
 - unless e.g., we have unique values per row (e.g., if a participant gave a free-text answer)
 - or perhaps we have stored some stimuli sentences
 - * although this would arguably be better as a 'category', since there should be multiple trials across participants that contain the same sentences

numeric class

- variables with numeric values, usually some variable we'd want to compute summaries on, e.g., means
- sometimes we don't want numbers to be stored as numeric class, however
 - this is the case for our variables `yes_press` and `KeyPress` (with 4 or 5)
- the same is true for our variable `item_id`, which ranges from 1:120
 - the numbers are just unique codes for our stimuli, the difference between `item 1` and `item 2` has nothing to do with the difference between the numbers 1 and 2

factor class

- we typically want *grouping* variables to be **factor** class
 - factors contain *categorical* data
 - any number that could be replaced with some other label should be a factor
- region of interest (ROI) = 1:7
 - but we want to know how many observations per region, the number is not informative
 - ROI could alternatively be coded as, e.g., "adverb", "pronoun", "verb", "spillover"

factor class

- let's change `df_lifetime$yes_press` to factor
 - using the `mutate()` verb from `dplyr`
 - and `as_factor()` from `forcats`

```
1 # change yes_press to factor
2 df_lifetime %>%
3   mutate(yes_press = as_factor(yes_press)) %>%
4   summary()
```

px	trial	eye	IA_DWELL_TIME
Length:4431	Min. : 1.0	Length:4431	Min. : 0.0
Class :character	1st Qu.: 52.5	Class :character	1st Qu.: 0.0
Mode :character	Median :104.0	Mode :character	Median : 301.0
	Mean :105.0		Mean : 587.5
	3rd Qu.:157.0		3rd Qu.: 765.5
	Max. :208.0		Max. :8968.0
IA_FIRST_FIXATION_DURATION	IA_FIRST_RUN_DWELL_TIME	IA_FIXATION_COUNT	
Min. : 0.0	Min. : 0.0	Min. : 0.000	
1st Qu.: 0.0	1st Qu.: 0.0	1st Qu.: 0.000	
Median :161.0	Median : 245.0	Median : 2.000	
Mean :139.4	Mean : 507.9	Mean : 2.714	
3rd Qu.:202.5	3rd Qu.: 586.0	3rd Qu.: 4.000	
Max. :775.0	Max. :8968.0	Max. :35.000	
IA_ID	IA_LABEL	IA_REGRESSION_IN	IA_REGRESSION_IN_COUNT
Min. :1.000	Length:4431	Min. :0.00000	Min. :0.0000
1st Qu.:1.000	Class :character	1st Qu.:0.00000	1st Qu.:0.0000
Median :2.000	Mode :character	Median :0.00000	Median :0.0000
Mean :2.681		Mean :0.09817	Mean :0.1318
3rd Qu.:4.000		3rd Qu.:0.00000	3rd Qu.:0.0000
Max. :6.000		Max. :1.00000	Max. :5.0000
IA_REGRESSION_OUT	IA_REGRESSION_OUT_COUNT	IA_REGRESSION_PATH_DURATION	
Min. :0.00000	Min. :0.00000	Min. : 0.0	
1st Qu.:0.00000	1st Qu.:0.00000	1st Qu.: 0.0	
Median :0.00000	Median :0.00000	Median : 282.0	
Mean :0.08147	Mean :0.09185	Mean : 595.6	
3rd Qu.:0.00000	3rd Qu.:0.00000	3rd Qu.: 747.0	
Max. :1.00000	Max. :7.00000	Max. :10242.0	
KeyPress	rt	bio	critical
Min. :4.000	Min. : 533	Length:4431	Length:4431


```

1st Qu.:4.000  1st Qu.: 1332  Class :character  Class :character
Median :4.000  Median : 1890  Mode  :character  Mode  :character
Mean   :4.496  Mean   : 2467
3rd Qu.:5.000  3rd Qu.: 2910
Max.   :5.000  Max.   :15654

  gender      item_id      list      match
Length:4431   Min.   :  1.00   Min.   :14.00   Length:4431
Class :character 1st Qu.: 26.00   1st Qu.:15.00   Class :character
Mode  :character Median : 51.00   Median :25.00   Mode  :character
              Mean   : 64.16   Mean   :29.45
              3rd Qu.: 78.50   3rd Qu.:35.00
              Max.   :208.00   Max.   :45.00

  condition      name      name_vital_status      tense
Length:4431   Length:4431   Length:4431   Length:4431
Class :character Class :character Class :character Class :character
Mode  :character Mode  :character Mode  :character Mode  :character

  type      yes_press
Length:4431  4:2218
Class :character 5:2213
Mode  :character

```

multiple arguments in a verb

- we can also change multiple columns at once:

```

# change ROI & label to factor
df_lifetime %>%
  mutate(KeyPress = as_factor(KeyPress),
         item_id = as_factor(item_id)) %>%
  summary()

```

```

  px      trial      eye      IA_DWELL_TIME
Length:4431   Min.   :  1.0   Length:4431   Min.   :  0.0
Class :character 1st Qu.: 52.5   Class :character 1st Qu.:  0.0
Mode  :character Median :104.0   Mode  :character Median : 301.0

```

Mean	:105.0	Mean	: 587.5
3rd Qu.:	157.0	3rd Qu.:	765.5
Max.	:208.0	Max.	:8968.0

IA_FIRST_FIXATION_DURATION	IA_FIRST_RUN_DWELL_TIME	IA_FIXATION_COUNT
Min. : 0.0	Min. : 0.0	Min. : 0.000
1st Qu.: 0.0	1st Qu.: 0.0	1st Qu.: 0.000
Median :161.0	Median : 245.0	Median : 2.000
Mean :139.4	Mean : 507.9	Mean : 2.714
3rd Qu.:202.5	3rd Qu.: 586.0	3rd Qu.: 4.000
Max. :775.0	Max. :8968.0	Max. :35.000

IA_ID	IA_LABEL	IA_REGRESSION_IN	IA_REGRESSION_IN_COUNT
Min. :1.000	Length:4431	Min. :0.00000	Min. :0.0000
1st Qu.:1.000	Class :character	1st Qu.:0.00000	1st Qu.:0.0000
Median :2.000	Mode :character	Median :0.00000	Median :0.0000
Mean :2.681		Mean :0.09817	Mean :0.1318
3rd Qu.:4.000		3rd Qu.:0.00000	3rd Qu.:0.0000
Max. :6.000		Max. :1.00000	Max. :5.0000

IA_REGRESSION_OUT	IA_REGRESSION_OUT_COUNT	IA_REGRESSION_PATH_DURATION	KeyPress
Min. :0.00000	Min. :0.00000	Min. : 0.0	4:2234
1st Qu.:0.00000	1st Qu.:0.00000	1st Qu.: 0.0	5:2197
Median :0.00000	Median :0.00000	Median : 282.0	
Mean :0.08147	Mean :0.09185	Mean : 595.6	
3rd Qu.:0.00000	3rd Qu.:0.00000	3rd Qu.: 747.0	
Max. :1.00000	Max. :7.00000	Max. :10242.0	

rt	bio	critical	gender
Min. : 533	Length:4431	Length:4431	Length:4431
1st Qu.: 1332	Class :character	Class :character	Class :character
Median : 1890	Mode :character	Mode :character	Mode :character
Mean : 2467			
3rd Qu.: 2910			
Max. :15654			

item_id	list	match	condition
2 : 48	Min. :14.00	Length:4431	Length:4431
7 : 48	1st Qu.:15.00	Class :character	Class :character
8 : 48	Median :25.00	Mode :character	Mode :character
9 : 48	Mean :29.45		
10 : 48	3rd Qu.:35.00		
12 : 48	Max. :45.00		

(Other):4143

name	name_vital_status	tense	type
Length:4431	Length:4431	Length:4431	Length:4431
Class :character	Class :character	Class :character	Class :character
Mode :character	Mode :character	Mode :character	Mode :character

yes_press
Min. :4.000
1st Qu.:4.000
Median :4.000
Mean :4.499
3rd Qu.:5.000
Max. :5.000

Pop quiz

1. Which class *should* the following variables be (numeric, factor, or character)?:

- participant ID
- trial number
- first-pass reading time
- regression path duration
- regressions in
- context sentence
- lifetime
- tense
- celebrity name

2. change them to these class types, and print a summary

3. save and render the document

Plot the data

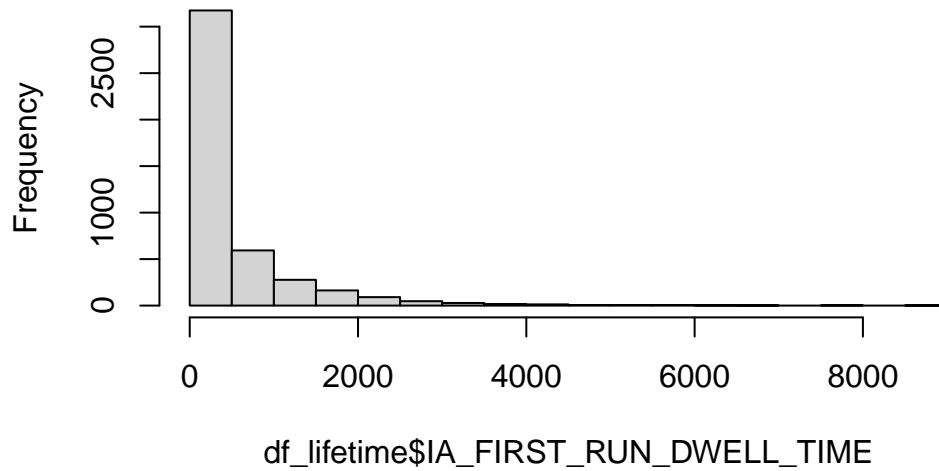
- at this stage we want to explore the data
 - distribution
 - * peaks, spread

– boundaries

Histogram

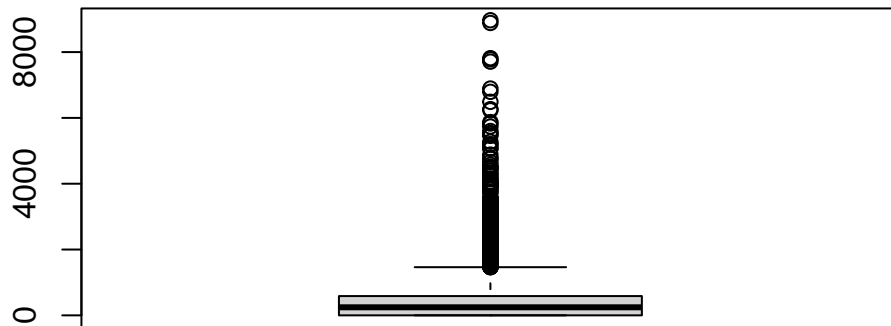
```
hist(df_lifetime$IA_FIRST_RUN_DWELL_TIME)
```

Histogram of df_lifetime\$IA_FIRST_RUN_DWELL_TIME



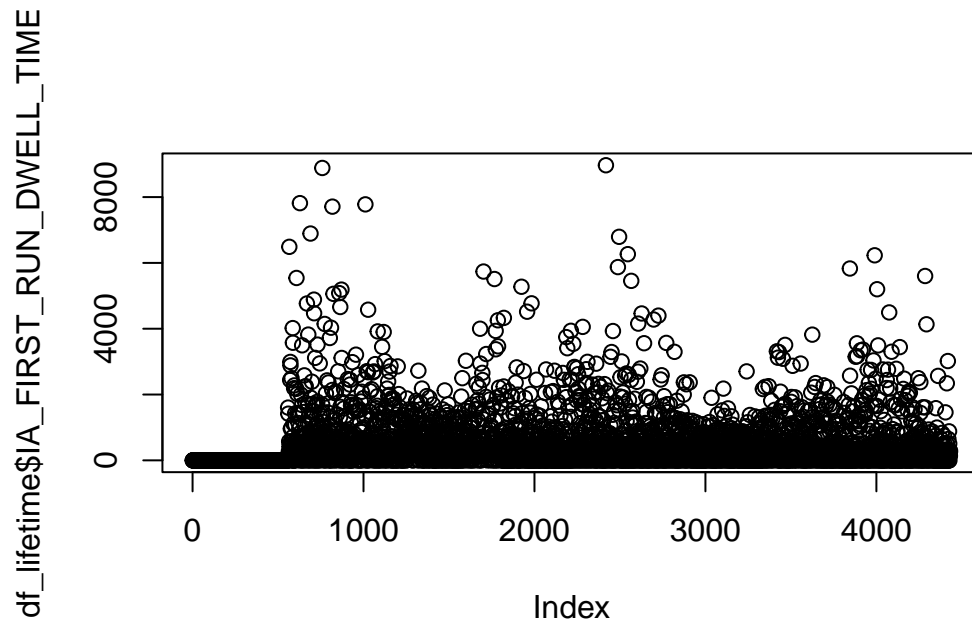
Boxplot

```
boxplot(df_lifetime$IA_FIRST_RUN_DWELL_TIME)
```



Scatterplot

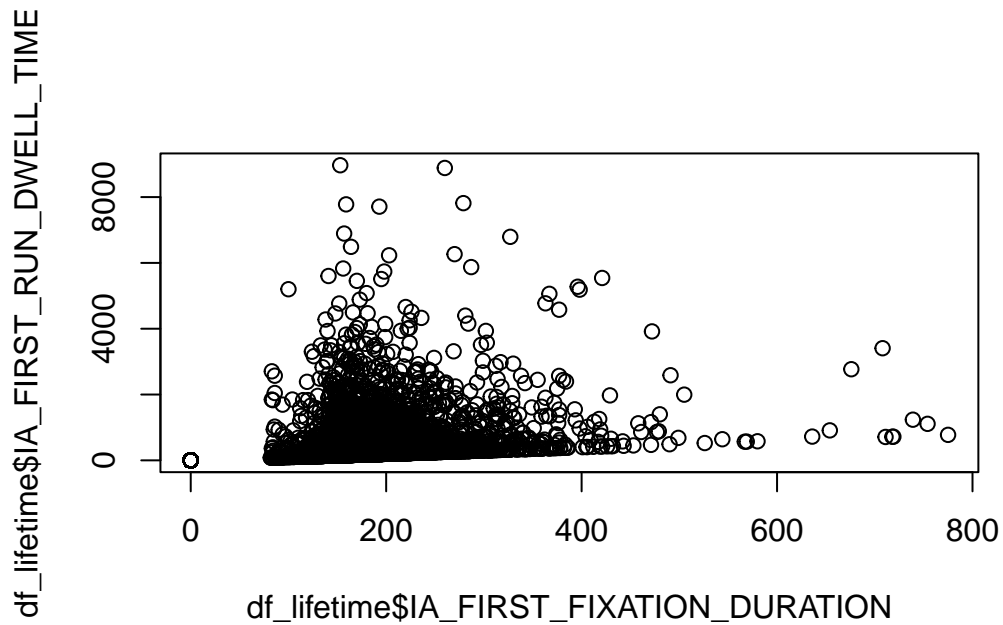
```
plot(df_lifetime$IA_FIRST_RUN_DWELL_TIME)
```



Plotting two variables

Scatterplot

```
plot(df_lifetime$IA_FIRST_FIXATION_DURATION, df_lifetime$IA_FIRST_RUN_DWELL_TIME)
```



Exercise

In your Quarto document:

1. create a heading 'Data exploration'
 - briefly describe the data
2. For each of our dependent variables:
 - create a subheading
 - calculate the mean and standard deviation of the variable (`mean()`, `sd()`) + create a boxplot of the variable
3. Render the document often to make sure it runs
4. Upload the source file (day1-nachname_vorname.qmd) to Moodle
5. download the source file below yours in the list to the same folder, and try to run it
 - does it run?

💡 print options

- each code chunk can have different print options:
 - `eval = FALSE`: do not evaluate this chunk
 - `include = FALSE` evaluate this chunk but don't show it or its results
 - `echo = FALSE` print this chunk code
 - `message = FALSE/warning = false` don't print warnings or messages
 - `error = TRUE` continue rendering document even if there's an error
 - * do not use `error = TRUE` for final versions! You want to make sure things work as they should

```
```{r, eval = T, echo = T, results = "asis", warning}  
code here
```
```

or

```
```{r}  
#| eval: false
code here
```
```

Session Info

```
sessionInfo()
```

```
R version 4.2.3 (2023-03-15)  
Platform: aarch64-apple-darwin20 (64-bit)  
Running under: macOS Ventura 13.2.1
```

```
Matrix products: default
```

```
BLAS: /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/lib/libRblas.0.dylib
```

```
LAPACK: /Library/Frameworks/R.framework/Versions/4.2-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
```


attached base packages:

```
[1] stats      graphics  grDevices  utils      datasets  methods   base
```

other attached packages:

```
[1] rbbt_0.0.0.9000 beeper_1.3      lubridate_1.9.2 forcats_1.0.0  
[5] stringr_1.5.0   dplyr_1.1.1     purrr_1.0.1    readr_2.1.4  
[9] tidyr_1.3.0     tibble_3.2.1    ggplot2_3.4.2  tidyverse_2.0.0  
[13] here_1.0.1
```

loaded via a namespace (and not attached):

```
[1] pillar_1.9.0      compiler_4.2.3  tools_4.2.3     bit_4.0.5  
[5] digest_0.6.31     timechange_0.2.0 jsonlite_1.8.4  evaluate_0.20  
[9] lifecycle_1.0.3   gtable_0.3.3    pkgconfig_2.0.3 rlang_1.1.0  
[13] cli_3.6.1         rstudioapi_0.14 parallel_4.2.3  curl_5.0.0  
[17] yaml_2.3.7        xfun_0.38       fastmap_1.1.1   httr_1.4.5  
[21] withr_2.5.0       knitr_1.42      fs_1.6.1        generics_0.1.3  
[25] vctrs_0.6.1       hms_1.1.3       bit64_4.0.5     rprojroot_2.0.3  
[29] grid_4.2.3        tidyselect_1.2.0 glue_1.6.2      R6_2.5.1  
[33] fansi_1.0.4       vroom_1.6.1     rmarkdown_2.21  tzdb_0.3.0  
[37] magrittr_2.0.3    scales_1.2.1    htmltools_0.5.5 colorspace_2.1-0  
[41] utf8_1.2.3        stringi_1.7.12  munsell_0.5.0   crayon_1.5.2  
[45] audio_0.1-10
```

References

- Comrie, B. (1976). *Aspect: An introduction to the study of verbal aspect and related problems*. Cambridge: Cambridge University Press.
- Mittwoch, A. (2008a). Tenses for the living and the dead. *Theoretical and Crosslinguistic Approaches to the Semantics of Aspect*, 110, 167.
- Mittwoch, A. (2008b). The English Resultative perfect and its relationship to the Experiential perfect and the simple past tense. *Linguistics and Philosophy*, 31(3), 323–351. <https://doi.org/10.1007/s10988-008-9037-y>