

R workshop

Data manupulation & Rmarkdown

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About me



- Statistician at NEAR, KI.
- MSc in Statistics, Uppsala University.
- Interests:
 - R package dev
 - Statistics
 - Data manipulation
 - guitar

Photo: Maria Yohuang

workshopr package Install

```
install.packages("remotes")
remotes::install_github("Bolin-Wu/workshopr",
    subdir = "rpackage",
    force = TRUE
)
```

Load

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

Data manupulation

tidyverse, assign

Introduction

This session is to share useful data manipulation skills at daily data harmonization work. My main goal is to follow the "don't repeat yourself" (DRY) principle.

It can make our code more readable and reduce our chance of making mistakes.

Suppose you want to change column class.

```
df$cohort1_edu = as.factor(df$cohort1_edu )
df$cohort2_edu = as.factor(df$cohort2_edu)
# and so on...
```

Get the code

Note: Data frames may seem to be unfit in the PDF. I do not user any html widges since the code will be pulled out for tutorial purpose. Attendants can run the code on their on machine instead to get better view.

```
workshopr::get_code_2023(session = "tidyverse")
```

Content

The content is selected based on data manipulation in real work scenario.

I hope by the end of the workshop, you will have them in your toolbox:

- %>% syntax
- join data frames (join function)
- transform data shape (pivot_longer)
- select variables based on name pattern (select)
- extract the label from DTA and SPSS in R (filter & sjlabelled)
- check missing values (summarise & across)
- mutate data based on column types (mutate & across)
- bin variables by percentiles (case_when)
- assign function



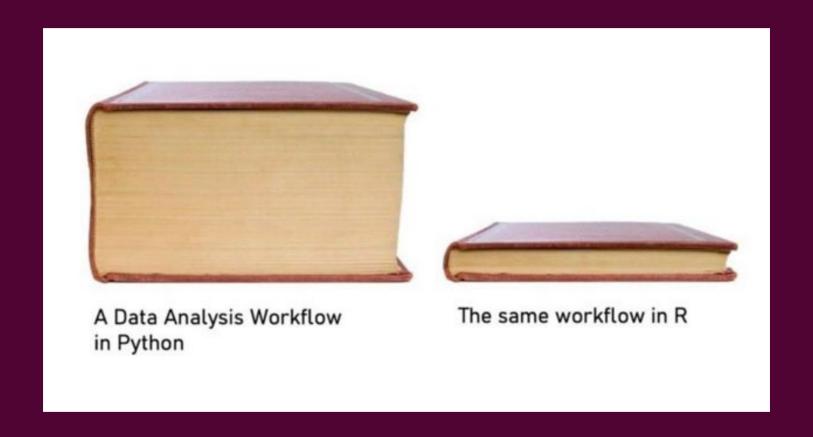
Tidyverse

Many times we just library (tidyverse). Actually <u>Tidyverse</u> is a huge umbrella consists of several powerful visualization and data manipulation package. For example:

- magrittr: pipeline operator %>%.
- ggplot2: ggplot().
- dplyr: select(), filter(), mutate().
- stringr: str_detect(), str_subset().

Notes

- Advantage: All at once.
- Disadvantage:
 - Slower to load the whole package.
 - Potential conflicts of function names with other packages. Example here.



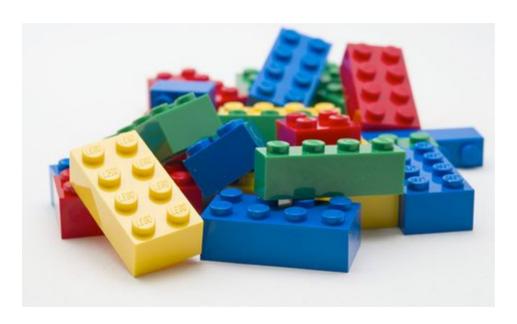
From Matt Dancho's Linkedin



picture credit: Wikipedia of Hadley Wickham

Pipeline operator %>%

Beautiful syntax with pipeline, just like playing LEGO.



```
fake_data %>%
   select(Lopnr, Date_wave1)
# A tibble: 3,000 \times 2
   Lopnr Date_wave1
   <dbl> <date>
       1 2001-06-11
       2 2002-04-08
       3 2001-09-12
 4
       4 2001-06-13
 5
       5 2001-02-12
 6
       6 2001-07-05
       7 2001-12-28
 8
       8 2002-09-17
9
       9 2001-09-14
10
   10 2001-09-21
# ... with 2,990 more rows
```

In addition, filter on the date column

4 15 2002-04-28

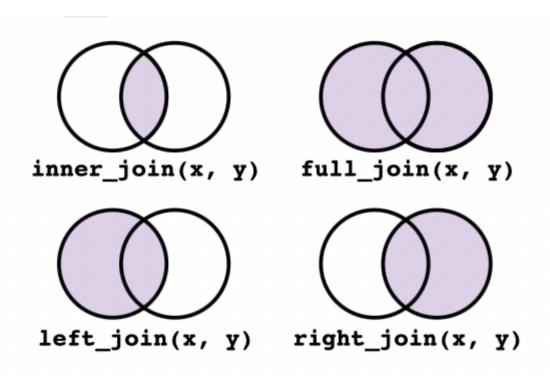
19 2002-12-25

5

You can further stacking group, filter, mutate, etc., with the pipeline operator.

Join data frames

■ With .*_join() function: There are 4 common types of joins.



■ Take left_join() as an example:

```
# From `dplyr` documentation:
df1 <- tibble(x = 1:3)
df2 <- tibble(
    x = c(1, 1, 2),
    y = c("first", "second", "third")
)
df1 %>% left_join(df2)
```

By the way, do you know why tibble() is better than data.frame()?

Transform data shape

Transform data shape is frequently used to clean data. However, for many people, including me, it sounds troublesome. In R, its relevant functions are evolving overtime as well.

In the beginning (2019), I used spread() and gather(). Every time I use spread() and gather(), it takes me a while to figure out how to fill in 'key' and 'value'. But as you can see from their documentation, their 'lifecycle' is 'superseded'.

Transform data shape

Now I only use pivot_longer() and pivot_wider() for transforming data. You can find their comprehensive documentation here.

They come with better documentation, more powerful application, and better integration with tidyverse syntax.

Let's assume we received a wide format data:

```
head(fake_data, n = 5)
# A tibble: 5 \times 28
 Lopnr Date_wave1 age_wave1 mmse_...¹ demen...² Date_wave2 age_w
 <db
                                                  79
    1 2001-06-11 77.6
                             29
                                    0 2003-06-03
2 2 2002-04-08 72.6
                                                  74
                            18
                                    0 2004-03-07
3
  3 2001-09-12 79.9 22
                                                  NA
                                    0 NA
                                                  81
4 4 2001-06-13 79.9 26 0 2003-05-12
5
  5 2001-02-12 82.6 NA 0 2003-04-20
                                                  84
# ... with 21 more variables: mmse_wave2 <dbl>,
#
   dementia_wave2 <dbl>, Date_wave3 <date>, age_wave3 <dbl>,
#
   mmse_wave3 <dbl>, dementia_wave3 <dbl>, Date_wave4 <date>
#
   age_wave4 <dbl>, mmse_wave4 <dbl>, dementia_wave4 <dbl>,
#
   Date_wave5 <date>, age_wave5 <dbl>, mmse_wave5 <dbl>,
#
   dementia_wave5 <dbl>, Date_wave6 <date>, age_wave6 <dbl>,
#
   mmse_wave6 <dbl>, dementia_wave6 <dbl>, age_base <dbl>2...
```

The column names are:

```
sort(colnames(fake_data))
                      "age_wave1"
 [1] "age_base"
                                        "age_wave2"
 [4] "age_wave3"
                      "age_wave4"
                                        "age_wave5"
 [7] "age_wave6"
                      "Date_wave1"
                                        "Date_wave2"
[10] "Date_wave3"
                      "Date_wave4"
                                        "Date_wave5"
[13] "Date_wave6"
                      "dementia_wave1" "dementia_wave2"
[16] "dementia_wave3" "dementia_wave4" "dementia_wave5"
[19] "dementia_wave6" "education"
                                        "Lopnr"
[22] "mmse_wave1"
                                        "mmse_wave3"
                      "mmse_wave2"
[25] "mmse_wave4"
                                        "mmse_wave6"
                      "mmse_wave5"
[28] "sex"
```

Now, assume for some reason, e.g. merge it with other data set, we want to transform it in a long format.

There are 3 variables with prefix should be formatted: 'age', $^{28/99}$

Select the interested columns

```
fake_data %>%
  select(contains("Date")) %>%
  slice(1:5)
# A tibble: 5 \times 6
  Date wave1 Date wave2 Date wave3 Date wave4 Date wave5
  <date> <date> <date> <date>
1 2001-06-11 2003-06-03 2005-05-27 2007-04-22 2009-04-10
2 2002-04-08 2004-03-07 2006-02-08 2008-02-19 2010-03-26
3 2001-09-12 NA
                       NA
                                  NA
                                             NA
4 2001-06-13 2003-05-12 2005-04-20 2007-03-28 2009-04-10
5 2001-02-12 2003-04-20 2005-04-06 2007-03-14 2009-03-17
# ... with 1 more variable: Date wave6 <date>
```

Read documentation, try to fill in the arguments.

```
?tidyr::pivot_longer()
```

- The 3 basic arguments are:
 - cols(): tells R what variables to pivot.
 - names_to(): a new name for columns in cols().
 - values_to(): a new name for values under the columns in cols().

Let's give a first try:

```
fake_data %>%
  select(Lopnr, contains("Date")) %>%
  pivot_longer(
    cols = contains("Date"),
    names_to = "wave", values_to = "date",
    names_prefix = "Date"
)
```

The result above looks good, but 'wave' looks a bit strange. I will leave the task to audience to fix this column.

Do the same with 'dementia' columns

```
fake_data %>%
  select(Lopnr, contains("dementia")) %>%
  pivot_longer(
    cols = contains("dementia"),
    names_to = "wave", names_prefix = "dementia",
    values_to = "dementia"
)
```

Data transform exercise (5 - 10 min)

 Read documentation. Change the arguments in the pivot_longer() function to get proper wave column.

Example result (first 5 rows):

Merge the two long pivot data sets together.

Consider: is it enough to only join on one column? Why or why not?

Select variables by name pattern

Some times you get data with specific variables. E.g.

- Each wave has its own specific prefix.
- Questionnaire/in person test has its own prefix.
- **...**

In this case, the select() function can help you. Some useful **selection helpers** are

```
starts_with()
```

- contains()
- matches()
- **...**

More details please see documentation:

```
?select()
```

starts_with

```
fake_data %>%
  select(starts_with("Date"))
```

contains

```
fake_data %>%
  select(contains("Date") & contains("1"))
```

If we want to select variables containing 'Date' or 'age', how to do it? Let's try it.

match

- match() is more advanced since it uses regular expression (regex).
- There are many regex cheatsheets online. For exmaple this.
- One useful website to test regex is here.

```
fake_data %>%
  select(matches("Date_wave\\d"))
```

---> Live demo

Get labels from datasets

- When we get SPSS or STATA data sets, usually they come with labels. E.g. in SPSS, one can check them in the "Variable View" tab.
- In R, one can use view() function. In the example data, we have:



A natural question to ask: how to extract the labels in R? If you run str() function, the labels are in the "label" attribute. There are multiple ways to extract the labels.

```
str(fake_data)
```

The one I use is the sjlabelled R package.

```
label_char <- sjlabelled::get_label(fake_data)
label_char</pre>
```

```
Date_wave1
                        Lopnr
                         "ID"
                               "date examination 2011-2012"
                    age_wave1
                                                  mmse_wave1
                                   "mmse at visit 2011-2012"
     "age at visit 2011-2012"
               dementia_wave1
                                                  Date_wave2
                               "date examination 2011-2012"
"dementia at visit 2011-2012"
                                                  mmse_wave2
                    age_wave2
     "age at visit 2011-2012"
                                   "mmse at visit 2011-2012"
               dementia_wave2
                                                  Date_wave3
                               "date examination 2011-2012"
"dementia at visit 2011-2012"
```

The result looks terrible, so we have to fix it by transforming it to be tibble:

```
label_df <- tibble::rownames_to_column(</pre>
   as.data.frame(label_char),
   "variable"
label_df <- tibble::as_tibble(label_df)</pre>
label_df
# A tibble: 28 \times 2
   variable
                  label_char
   <chr>>
                  <chr>
                  ID
 1 Lopnr
                  date examination 2011-2012
 2 Date wave1
 3 age_wave1
                  age at visit 2011-2012
                  mmse at visit 2011-2012
 4 mmse_wave1
 5 dementia_wavel dementia at visit 2011-2012
                  date examination 2011-2012
 6 Date_wave2
                  age at visit 2011-2012
 7 age_wave2
                                                       50 / 99
```

mmse at visit 2011-2012

8 mmse_wave2

The result looks much better! Now we can do lots of things with pipeline.

filter the label contains 'dementia'

```
label_df %>%
  filter(grepl("dementia", label_char))
# A tibble: 6 \times 2
 variable label_char
 <chr>
1 dementia_wavel dementia at visit 2011-2012
2 dementia wave2 dementia at visit 2011-2012
3 dementia wave3 dementia at visit 2011-2012
4 dementia_wave4 dementia at visit 2011-2012
5 dementia wave5 dementia at visit 2011-2012
6 dementia_wave6 dementia at visit 2011-2012
```

filter the variable contains 'wave'

```
label_df %>%
  filter(grepl("wave", variable))
# A tibble: 24 \times 2
  variable label char
  <chr>
          <chr>
1 Date_wave1 date examination 2011-2012
 2 age_wave1 age at visit 2011-2012
 3 mmse_wave1 mmse at visit 2011-2012
4 dementia wavel dementia at visit 2011-2012
 5 Date_wave2 date examination 2011-2012
6 age_wave2 age at visit 2011-2012
7 mmse_wave2 mmse at visit 2011-2012
 8 dementia_wave2 dementia at visit 2011-2012
9 Date_wave3 date examination 2011-2012
10 age_wave3 age at visit 2011-2012
# ... with 14 more rows
```

Missing values Count the NA

■ If count the NA of one column. It could something like:

```
sum(is.na(fake_data$Date_wave1))
```

[1] 0

What if multiple columns?

Count NA in multiple columns

```
fake_data %>%
  summarise(across(
    where(lubridate::is.Date),
    ~ sum(is.na(.))
  ))
```

- You may wonder: what is this ~ sum(is.na(.))?
- It is a purrr-style lambda function.
- You may also wonderL what is across?
- It tells summarise to do what operations, on which

Missing value exercise (5 - 10 min)

- Read the documentation of across() function.
- Count the NA of all numeric/int columns.
- Count the NA of columns contains certain strings, e.g. 'edu'

Example result

```
# A tibble: 1 \times 22
  Lopnr age_w...¹ mmse_...² demen...³ age_w...⁴ mmse_...⁵ demen...6 age_w
  <int> <int> <int> <int> <int> <int> <int>
                                                           < in
                    135
                                    207
                                             323
                                                     207
# ... with 14 more variables: mmse_wave3 <int>,
#
    dementia_wave3 <int>, age_wave4 <int>, mmse_wave4 <int>,
    dementia_wave4 <int>, age_wave5 <int>, mmse_wave5 <int>,
#
    dementia_wave5 <int>, age_wave6 <int>, mmse_wave6 <int>,
#
#
    dementia_wave6 <int>, age_base <int>, sex <int>,
    education <int>, and abbreviated variable names
```

Mutation based on column type

- Now you should have some understanding of where() function.
- When you apply it with mutate() function, you can do many manipulations. E.g. round the digits of numeric columns.

```
fake_data %>%
  mutate(across(
    where(is.numeric),
    ~ round(., digits = 2)
  ))
```

A tibble: 3,000 × 28

Bin variables

- case_when() function can do the work for us easily, it accommodates well with pipeline syntax.
- For example, if we want to bin the education variable.

```
fake_data %>%
  transmute(mmse_wave1,
    mmse_wave1_bin = case_when(
       between(mmse_wave1, 0, 9) ~ "Severe dementia",
       between(mmse_wave1, 10, 18) ~ "Moderate dementiant
       between(mmse_wave1, 19, 23) ~ "Mild dementiant",
       mmse_wave1 >= 24 ~ "Mild dementiant",
       TRUE ~ NA_character_
    )
)
```

Assign function

I would like to spend some time introduce assign() function in R.

- Even though it is not part of tidyverse family. But it is really useful! For example, when you bulk import databases into R environment; or you want to bulk change the imported data frame names.
- A simple example (run in your R studio):

```
df_names <- c()
set.seed(2023)
for (i in 1:4) {
   df_names[i] <- paste0("edu_", i)
   assign(df_names[i], sample.int(3, 10, replace = T)
   # print the result
   cat(
    "The df name is: ", df_names[i], "\n",
    "its value is: ", get(df_names[i]), "\n"</pre>
```

Messy dataframe name

- Imagine in your environment you have these dataframe.
 - Chances are that they are named this way when you get the data files from someone. They have spaces, "-", horrible!

```
objects_name <- c(
           "Cohort1_Baseline_BMI", "Cohort1_FU1_BMI", "Cohort1
           "Cohort1FU3_Cohort2FU2", "Cohort2_Baseline_BMI", "
           "Gender data request_20230111", "Gender data request
           "index", "NEAR_BMI-mortality", "SNAC-K C1 B-F6 cohe
           "SNAC-K C1_B", "SNAC-K C1_F1", "SNAC-K C1_F2", "SNAC-K C1_F2",
           "SNAC-K C1_F4", "SNAC-K C1_F5"
for (i in 1:length(objects_name)) {
           # assign some values to these objects
           assign(objects_name[i], sample(100, 10))
```

■ Look at your 'Environment' panel. Or ls() can retrieve all object names in your R environment. Do you feel familiar?

Let's fix it with assign() function (one possible way).

```
clean_names <- gsub(" |-", "_", objects_name)

for (i in 1:length(clean_names)) {
   assign(clean_names[i], get(objects_name[i]))
}

# just show the first few
clean_names[1:5]</pre>
```

- [1] "Cohort1_Baseline_BMI" "Cohort1_FU1_BMI"
 [3] "Cohort1_FU2_BMI" "Cohort1FU3_Cohort2FU2"
 [5] "Cohort2_Baseline_BMI"
 - Check your 'Environment' panel again.
 - The gsub() part can be customized with regex syntax. Basically you can fix any kind of messy dataframe names.

R markdown Basics and daily work uses

Introduction

- A complete Rmarkdown documentation could be found here.
- The key is to understand these three pillows:

Rmarkdown = markdown + yaml heading + code chunk options.

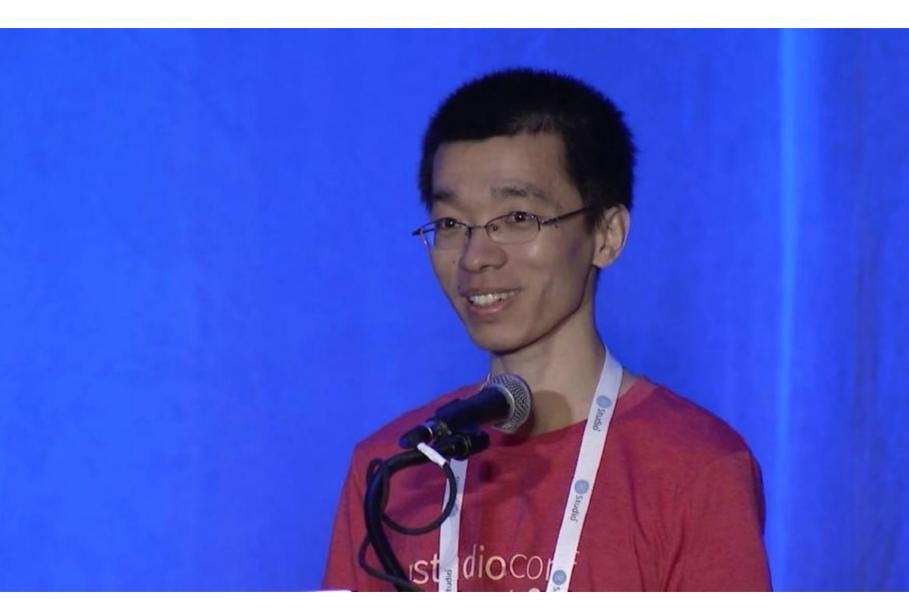
Markdown

- Markdownis a lightweight language for creating formatted text using a plain-text editor.*
- Advantage:
 - You do not need to worry about the format. Just focus on writing itself. Could be superfast.
 - Little chance of making the format consistent across the whole documentation.
- Disadvantage:
 - A steep learning curve in the beginning. * However, this disadvantage greatly remedied by many userfriendly editor, e.g. overleaf

Let me give a live example via an online markdown editor

code chunk options

- Code chunk options in RMD can help you control the execution of your code in the file.
- Some useful ones are as follows:*
- include = FALSE prevents code and results from appearing in the finished file. R Markdown still runs the code in the chunk, and the results can be used by other chunks.
- echo = FALSE prevents code, but not the results from appearing in the finished file. This is a useful way to embed figures.
- message = FALSE prevents messages that are generated by code from appearing in the finished file.
- warning = FALSE prevents warnings that are generated by code from appearing in the finished.



```
'``{r import package, message=FALSE}
library(tidyverse)
library(Hmisc)
'``
{r, include=FALSE}
library(DT)
'``
```

local option

```
# By default, all codes are hidden.
knitr::opts_chunk$set(message = FALSE, warning = FALSE, include = TRUE)
library(workshopr)
library(tidyverse)
library(DT)
```

RMD chunk option exercises (1-3 minutes)

- Print the column names of fake_data, but do not show the code itself.
- Print the code above, but do not show the result.
- Neither print the code nor the result, but preserve the evaluation of execution for further analysis.

YAML

```
title: "A Cool Presentation"
output: html_document
---
```

Or

```
title: "A Cool Presentation"
output:
   word_document:
    toc: yes
    toc_depth: 2
```

■ The YAML heading basically defines the type of output file, layout, etc.

Live demo in R studio

Get the templates of html or word output:

```
workshopr::get_rmd_2023(
   name = "pretty_template",
   output_file = "word"
)
workshopr::get_rmd_2023(
   name = "pretty_template",
   output_file = "html"
)
```

■ Let's take a look together!

Final exercise

Please check this link here.

Wrap up

- Many times, avoiding repetitive coding can help you, also others to review your code.
- In the beginning, the Rmarkdown has steep learning curve, but once you are familiar with it, your life will be easier.
- This whole slide is written in Rmarkdown. Source code is <u>here</u>. The R example code is pulled from Rmarkdown, I do not need to copy and paste.
- The materials are collected in Github repo: https://github.com/Bolin-Wu/workshopr
- Hope today's content could be an enlightment to you. If you have any question, please contact me: bolin.wu@ki.se;