

# Introduction to R

## 2.1 Data Wrangling - Tidyverse Philosophy

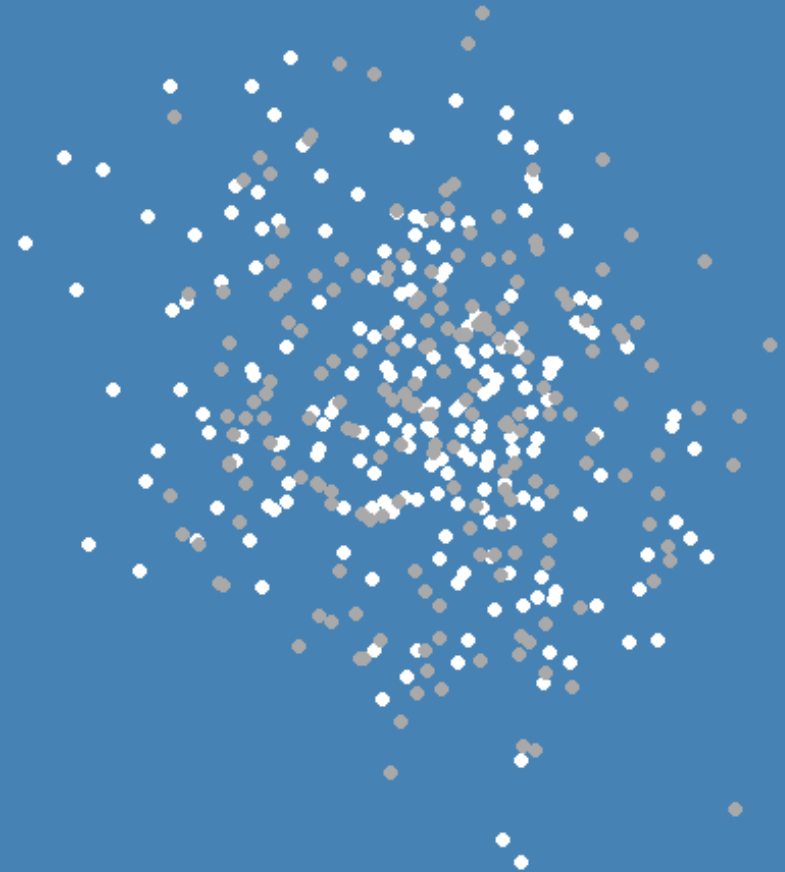
What is dplyr? Structure of a Function Call,  
Pipe Operator %>%

Lion Behrens, M.Sc.

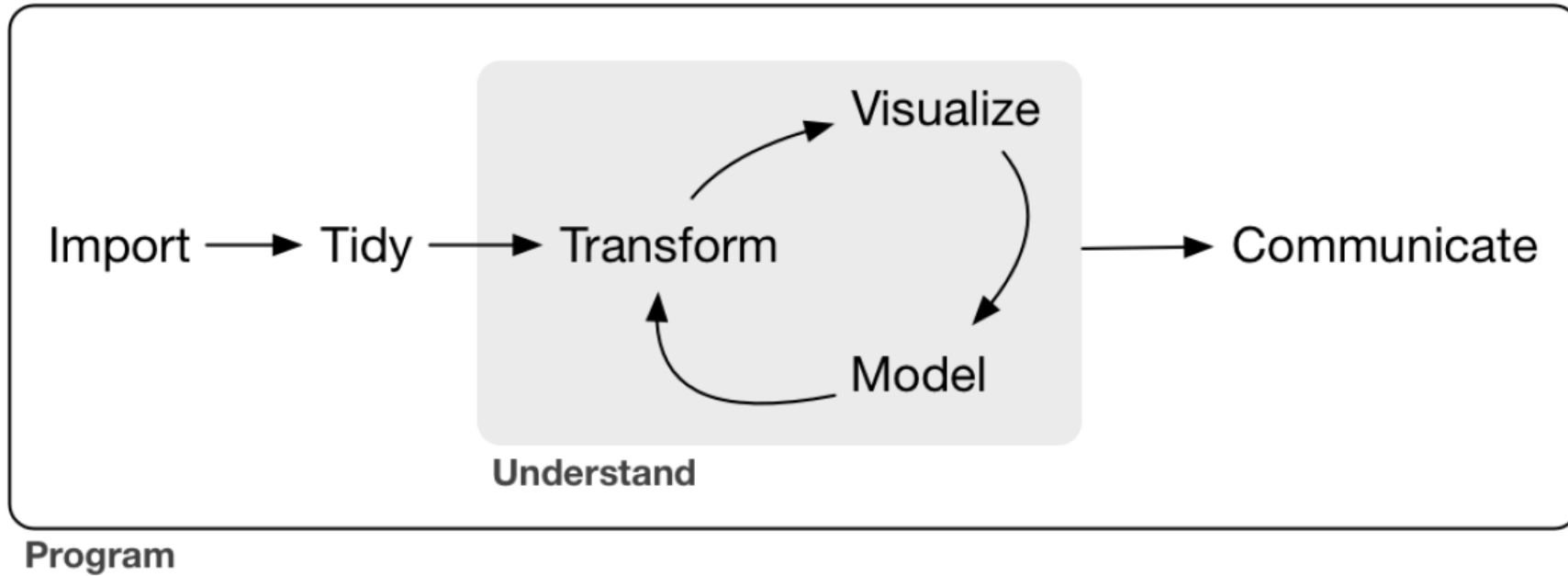


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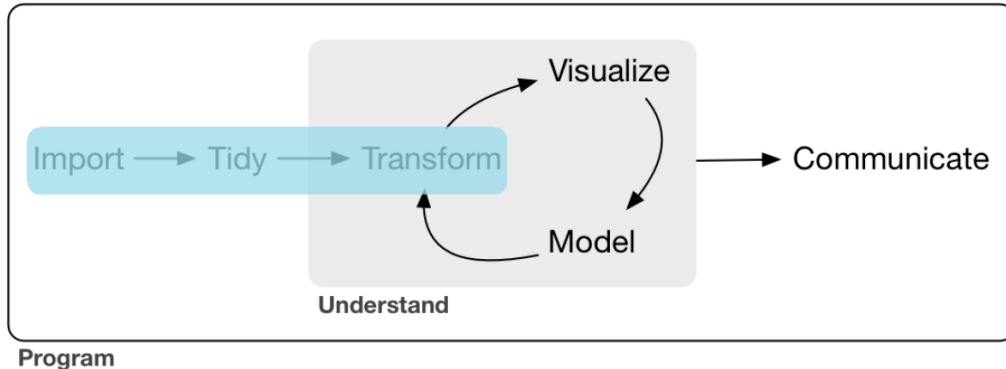


# Data Project Flow



Note: Figure from Wickham and Grolemund, 2017. R For Data Science. Sebastopol: O' REILLY.

# What is Data Wrangling?



Note: Figure adapted from Wickham and Grolemund, 2017. R For Data Science. Sebastopol: O' REILLY.

Data wrangling is the process of **getting your data into shape** so that you can **analyze, visualize** and **model** it. The most common steps are:

- **importing** data of various formats into R
- **renaming** variables
- **selecting** a subset of variables or **filtering** out a subset of cases
- **recoding** variables (defining missing values)
- **creating** new variables as transformations/mutations of existing variables

**Data wrangling** (or **feature engineering**) is one of the most important and time-consuming tasks in social science research.

# Tidyverse



## R packages for data science

The tidyverse is an opinionated **collection of R packages** designed for data science.

All packages **share** an underlying

- design philosophy
- grammar
- data structures

Install the complete tidyverse with

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

# Dplyr - A Language for Data Manipulation

A collection of **key function calls** provides the **verbs** for **dplyr's** language of data manipulation.



- **select()** subsets variables based on their names
- **filter()** subsets observations (rows) based on their values
- **relocate()** and **arrange()** re-order variables and observations
- **mutate()** generates new variables from existing variables
- **summarize()** reduces multiple values to single summaries
- **group\_by()** performs operations by group
- **left\_join()**, **right\_join()**, **full\_join()** (etc.) merge several data sets

... and there are many more!

# dplyr - Structur of a Function Call

## Example Data

```
library(readstata13)
ess10 <- read.dta13("dat/ess10.dta")
```

```
glimpse(ess10)
```

```
## Rows: 18,060
## Columns: 6
## $ name      [3m [38;5;246m<chr> [39m [23m "ESS10e01_2", "ESS10e01_2", "ESS10e01_2", "ESS10e0...
## $ essround   [3m [38;5;246m<dbl> [39m [23m 10, 10, 10, 10, 10, 10, 10, 10, 10, 10, 10, 10, 10...
## $ edition    [3m [38;5;246m<chr> [39m [23m "1.2", "1.2", "1.2", "1.2", "1.2", "1.2", "1.2", "...
## $ proddate   [3m [38;5;246m<chr> [39m [23m "28.06.2022", "28.06.2022", "28.06.2022", "28.06.2...
## $ idno       [3m [38;5;246m<dbl> [39m [23m 10002, 10006, 10009, 10024, 10027, 10048, 10053, 1...
## $ cntry      [3m [38;5;246m<chr> [39m [23m "BG", "BG", "BG", "BG", "BG", "BG", "BG", "BG", "B..."
```

# dplyr - Structur of a Function Call

Let's say we only want to **extract respondents** from the **country of Hungary**.

```
filter(  
  ess10,  
  cntry == "HU"  
)
```

- start with stating the verb (function): `filter()`

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- the first argument is the data frame that we are working with: `ess10`



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- start with stating the verb (function): `filter()`
- the first argument is the data frame that we are working with: `ess10`
- the second argument is the observations that we want to select: `cntry == "HU"`

# dplyr - Structur of a Function Call

Let's say we only want to **extract respondents** from the **country of Hungary**.

```
filter(  
  ess10,  
  cntry == "HU"  
)
```

- start with stating the verb (function): `filter()`
- the first argument is the data frame that we are working with: `ess10`
- the second argument is the observations that we want to select: `cntry == "HU"`
- we now have reduced our data frame to the  $n = 1,849$  respondents from Hungary

```
## Rows: 1,849  
## Columns: 6  
## $ name      [3m [38;5;246m<chr> [39m [23m "ESS10e01_2", "ESS10e01_2", "ESS10e01_2", "ESS10e0...  
## $ essround  [3m [38;5;246m<dbl> [39m [23m 10, 10, 10, 10, 10, 10, 10, 10, 10, 10, 10, 10, 10...  
## $ edition   [3m [38;5;246m<chr> [39m [23m "1.2", "1.2", "1.2", "1.2", "1.2", "1.2", "1.2", "...  
## $ proddate  [3m [38;5;246m<chr> [39m [23m "28.06.2022", "28.06.2022", "28.06.2022", "28.06.2...  
## $ idno      [3m [38;5;246m<dbl> [39m [23m 10002, 10006, 10017, 10018, 10027, 10043, 10045, 1...  
## $ cntry     [3m [38;5;246m<chr> [39m [23m "HU", "HU", "HU", "HU", "HU", "HU", "HU", "HU", "H...
```

# Tidyverse - Pipe Operator %>%

What if we want information on **respondents from Hungary** but are only interested in extracting the **design weight** (**dweight**) and **population weight** (**pweight**) that is attached to each respondent?

- This means we want to conduct **more than one operation** on a data set
- Enter: The **pipe operator** %>%

```
ess10 %>%  
  filter(cntry == "HU") %>%  
  select(dweight, pweight)
```

- start with the data frame that we are working with

# Tidyverse - Pipe Operator %>%

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- the second line filters for the observations that we want to select: **cntry == "HU"**

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- Enter: The **pipe operator** %>%

```
ess10 %>%  
  filter(cntry == "HU") %>%  
  select(dweight, pweight)
```

- start with the data frame that we are working with
- the second line filters for the observations that we want to select: **cntry == "HU"**
- the third line selects the variables we are interested in: **dweight, pweight**

Note: The pipe operator %>% is used for **passing information from one process to the other**.

# Tidyverse - Piping vs. Assigning

The pipe operator itself will **not store the output** in a new object.

```
# print the output, store nothing
```

```
ess10 %>%  
  filter(cntry == "HU") %>%  
  select(dweight, pweight)
```

```
# store the output, print nothing
```

```
weights_hu <- ess10 %>%  
  filter(cntry == "HU") %>%  
  select(dweight, pweight)
```

# Tidyverse Data - Tibble

There is basically a **new object type** that exists in the tidyverse! **Tibbles!**



## Tibble vs. Data Frames

- Tibbles are **modernized versions of data frames**
- Tibbles are designed to make **working** with data frames **smarter**
- `print(df)` only prints out first ten observations: output is **tidier**
- Additional metadata is displayed for each variable: output is **more informative**

# References

Parts of this course are inspired by the following resources:

- Wickham, Hadley and Garrett Grolemund, 2017. *R for Data Science - Import, Tidy, Transform, Visualize, and Model Data*. O'Reilly.
- Bahnsen, Oke and Guido Ropers, 2022. *Introduction to R for Quantitative Social Science*. Course held as part of the GESIS Workshop Series.
- Breuer, Johannes and Stefan Jünger, 2021. *Introduction to R for Data Analysis*. Course held as part of the GESIS Summer School in Survey Methodology.
- Teaching material developed by Verena Kunz, David Weyrauch, Oliver Rittmann and Viktoriia Semenova.