

# Grammar of data wrangling

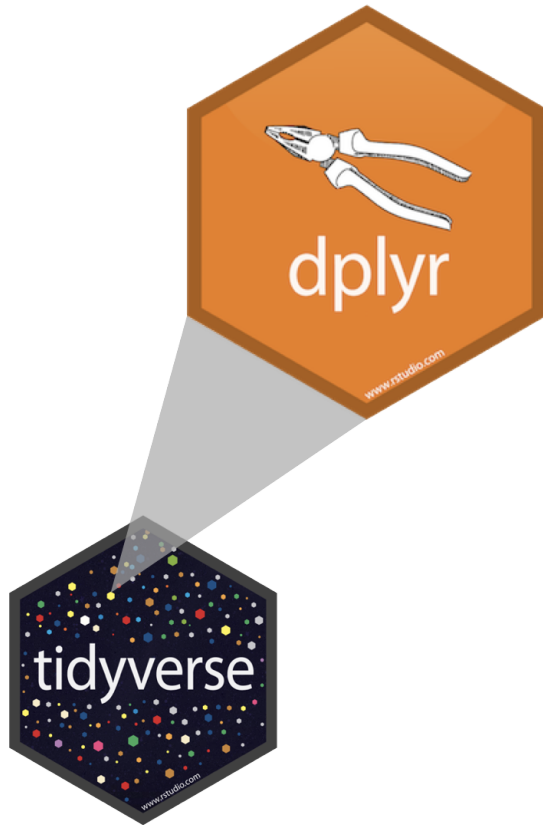
Learn how to transform data  
Prof. Dr. Jan Kirenz

*The following content is based on Mine Çetinkaya-Rundel's excellent material in "Data Science in a Box"*

# Grammar of data wrangling

# A grammar of data wrangling...

... based on the concepts of functions as verbs that manipulate data frames



- `select`: pick columns by name
- `arrange`: reorder rows
- `slice`: pick rows using index(es)
- `filter`: pick rows matching criteria
- `distinct`: filter for unique rows
- `mutate`: add new variables
- `summarise`: reduce variables to values
- `group_by`: for grouped operations
- ... (many more)

# Rules of `dplyr` functions

- First argument is *always* a data frame
- Subsequent arguments say what to do with that data frame
- Always return a data frame
- Don't modify in place

# Data: Hotel bookings

- Data from two hotels: one resort and one city hotel
- Observations: Each row represents a hotel booking
- Goal for original data collection: Development of prediction models to classify a hotel booking's likelihood to be cancelled ([Antonia et al., 2019](#))

```
hotels <- read_csv("data/hotels.csv")
```

Source: [TidyTuesday](#)

# First look: Variables

```
names(hotels)
```

```
## [1] "hotel"  
## [2] "is_canceled"  
## [3] "lead_time"  
## [4] "arrival_date_year"  
## [5] "arrival_date_month"  
## [6] "arrival_date_week_number"  
## [7] "arrival_date_day_of_month"  
## [8] "stays_in_weekend_nights"  
## [9] "stays_in_week_nights"  
## [10] "adults"  
## [11] "children"  
## [12] "babies"  
## [13] "meal"  
## [14] "country"  
## [15] "market_segment"  
## [16] "distribution_channel"  
## [17] "is_repeated_guest"  
## [18] "previous_cancellations"  
...
```

# Second look: Overview

```
glimpse(hotels)
```

```
## Rows: 119,390
## Columns: 32
## $ hotel                <chr> "Resort Hotel", "Resort...
## $ is_canceled          <dbl> 0, 0, 0, 0, 0, 0, 0, 0,...
## $ lead_time            <dbl> 342, 737, 7, 13, 14, 14...
## $ arrival_date_year    <dbl> 2015, 2015, 2015, 2015,...
## $ arrival_date_month   <chr> "July", "July", "July",...
## $ arrival_date_week_number <dbl> 27, 27, 27, 27, 27, 27,...
## $ arrival_date_day_of_month <dbl> 1, 1, 1, 1, 1, 1, 1, 1,...
## $ stays_in_weekend_nights <dbl> 0, 0, 0, 0, 0, 0, 0, 0,...
## $ stays_in_week_nights <dbl> 0, 0, 1, 1, 2, 2, 2, 2,...
## $ adults               <dbl> 2, 2, 1, 1, 2, 2, 2, 2,...
## $ children             <dbl> 0, 0, 0, 0, 0, 0, 0, 0,...
## $ babies               <dbl> 0, 0, 0, 0, 0, 0, 0, 0,...
## $ meal                 <chr> "BB", "BB", "BB", "BB",...
## $ country              <chr> "PRT", "PRT", "GBR", "G...
## $ market_segment       <chr> "Direct", "Direct", "Di...
## $ distribution_channel  <chr> "Direct", "Direct", "Di...
...

```

# Select a single column

View only `lead_time` (number of days between booking and arrival date):

```
select(hotels, lead_time)
```

```
## # A tibble: 119,390 x 1
##   lead_time
##   <dbl>
## 1      342
## 2      737
## 3         7
## 4        13
## 5        14
## 6        14
## # ... with 119,384 more rows
```



# Select a single column

```
select(  
  hotels,  
  lead_time  
)
```

- Start with the function (a verb):  
`select()`

# Select a single column

```
select(  
  hotels,  
  lead_time  
)
```

- Start with the function (a verb):  
`select()`
- First argument: data frame we're working with, `hotels`

# Select a single column

```
select(  
  hotels,  
  lead_time  
)
```

- Start with the function (a verb):  
`select()`
- First argument: data frame we're working with, `hotels`
- Second argument: variable we want to select, `lead_time`

# Select a single column

```
select(  
  hotels,  
  lead_time  
)
```

```
## # A tibble: 119,390 x 1  
##   lead_time  
##   <dbl>  
## 1      342  
## 2      737  
## 3         7  
## 4        13  
## 5        14  
## 6        14  
## # ... with 119,384 more rows
```

- Start with the function (a verb):  
`select()`
- First argument: data frame we're working with, `hotels`
- Second argument: variable we want to select, `lead_time`
- Result: data frame with 119390 rows and 1 column

dplyr functions always expect a data frame and always yield a data frame.

```
select(hotels, lead_time)
```

```
## # A tibble: 119,390 x 1
##   lead_time
##   <dbl>
## 1      342
## 2      737
## 3         7
## 4        13
## 5        14
## 6        14
## # ... with 119,384 more rows
```

# Select multiple columns

View only the `hotel` type and `lead_time`:

```
select(hotels, hotel, lead_time)
```

```
## # A tibble: 119,390 x 2
##   hotel      lead_time
##   <chr>      <dbl>
## 1 Resort Hotel      342
## 2 Resort Hotel      737
## 3 Resort Hotel       7
## 4 Resort Hotel      13
## 5 Resort Hotel      14
## 6 Resort Hotel      14
## # ... with 119,384 more rows
```

What if we wanted to select these columns, and then arrange the data in descending order of lead time?

# Data wrangling, step-by-step

Select:

```
hotels %>%  
  select(hotel, lead_time)
```

```
## # A tibble: 119,390 x 2  
##   hotel      lead_time  
##   <chr>      <dbl>  
## 1 Resort Hotel      342  
## 2 Resort Hotel     737  
## 3 Resort Hotel       7  
## 4 Resort Hotel     13  
## 5 Resort Hotel     14  
## 6 Resort Hotel     14  
## # ... with 119,384 more rows
```

Select, then arrange:

```
hotels %>%  
  select(hotel, lead_time) %>%  
  arrange(desc(lead_time))
```

```
## # A tibble: 119,390 x 2  
##   hotel      lead_time  
##   <chr>      <dbl>  
## 1 Resort Hotel     737  
## 2 Resort Hotel    709  
## 3 City Hotel      629  
## 4 City Hotel      629  
## 5 City Hotel      629  
## 6 City Hotel      629  
## # ... with 119,384 more rows
```

# Pipes



# What is a pipe?

In programming, a pipe is a technique for passing information from one process to another.

- Start with the data frame `hotels`, and pass it to the `select()` function,

```
hotels %>%  
  select(hotel, lead_time) %>%  
  arrange(desc(lead_time))
```

```
## # A tibble: 119,390 x 2  
##   hotel      lead_time  
##   <chr>      <dbl>  
## 1 Resort Hotel      737  
## 2 Resort Hotel      709  
## 3 City Hotel        629  
## 4 City Hotel        629  
## 5 City Hotel        629  
## 6 City Hotel        629  
## # ... with 119,384 more rows
```

# What is a pipe?

In programming, a pipe is a technique for passing information from one process to another.

- Start with the data frame `hotels`, and pass it to the `select()` function,
- then we select the variables `hotel` and `lead_time`,

```
hotels %>%  
  select(hotel, lead_time) %>%  
  arrange(desc(lead_time))
```

```
## # A tibble: 119,390 x 2  
##   hotel      lead_time  
##   <chr>      <dbl>  
## 1 Resort Hotel      737  
## 2 Resort Hotel      709  
## 3 City Hotel        629  
## 4 City Hotel        629  
## 5 City Hotel        629  
## 6 City Hotel        629  
## # ... with 119,384 more rows
```

# What is a pipe?

In programming, a pipe is a technique for passing information from one process to another.

- Start with the data frame `hotels`, and pass it to the `select()` function,
- then we select the variables `hotel` and `lead_time`,
- and then we arrange the data frame by `lead_time` in descending order.

```
hotels %>%  
  select(hotel, lead_time) %>%  
  arrange(desc(lead_time))
```

```
## # A tibble: 119,390 x 2  
##   hotel      lead_time  
##   <chr>      <dbl>  
## 1 Resort Hotel      737  
## 2 Resort Hotel      709  
## 3 City Hotel        629  
## 4 City Hotel        629  
## 5 City Hotel        629  
## 6 City Hotel        629  
## # ... with 119,384 more rows
```

# Aside

The pipe operator is implemented in the package **magrittr**, though we don't need to load this package explicitly since **tidyverse** does this for us.

Any guesses as to why the package is called magrittr?



# How does a pipe work?

- You can think about the following sequence of actions - find keys, unlock car, start car, drive to work, park.
- Expressed as a set of nested functions in R pseudocode this would look like:

```
park(drive(start_car(find("keys")), to = "work"))
```

- Writing it out using pipes give it a more natural (and easier to read) structure:

```
find("keys") %>%  
  start_car() %>%  
  drive(to = "work") %>%  
  park()
```

# A note on piping and layering

- `%>%` used mainly in **dplyr** pipelines, *we pipe the output of the previous line of code as the first input of the next line of code*
- `+` used in **ggplot2** plots is used for "layering", *we create the plot in layers, separated by +*

# dplyr



```
hotels +  
  select(hotel, lead_time)
```

```
## Error in select(hotel, lead_time): Objekt 'hotel' nicht gefunden
```



```
hotels %>%  
  select(hotel, lead_time)
```

```
## # A tibble: 119,390 x 2  
##   hotel      lead_time  
##   <chr>      <dbl>  
## 1 Resort Hotel      342  
## 2 Resort Hotel      737  
## 3 Resort Hotel        7  
## ...
```

# ggplot2

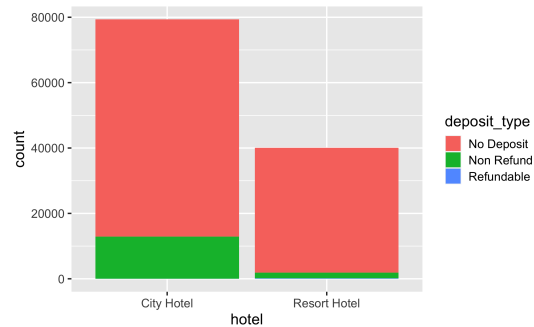


```
ggplot(hotels, aes(x = hotel, fill = deposit_type)) %>%  
  geom_bar()
```

```
## Error: `mapping` must be created by `aes()`  
## Did you use %>% instead of +?
```



```
ggplot(hotels, aes(x = hotel, fill = deposit_type)) +  
  geom_bar()
```





# Code styling

Many of the styling principles are consistent across `%>%` and `+`:

- always a space before
- always a line break after (for pipelines with more than 2 lines)



```
ggplot(hotels,aes(x=hotel,y=deposit_type))+geom_bar()
```



```
ggplot(hotels, aes(x = hotel, y = deposit_type)) +  
  geom_bar()
```