Filtering rows

ntro
Learning objectives
The Yaounde COVID-19 dataset
<pre>ntroducing filter()</pre>
Relational operators
Combining conditions with ${\tt \&}$ and ${\tt }$
Negating conditions with!
NA values
Wrap Up!

Intro

Onward with the {dplyr} package, discovering the filter verb. Last time we saw how to select variables (columns) and today we will see how to keep or drop data entries, rows, using filter. Dropping abnormal data entries or keeping subsets of your data points is another essential aspect of data wrangling.

Let's go!



Learning objectives

- 1. You can use dplyr::filter() to keep or drop rows from a dataframe.
- 2. You can filter rows by specifying conditions on numbers or strings using relational operators like greater than (>), less than (<), equal to (==), and not equal to (!=).
- 3. You can filter rows by combining conditions using logical operators like the ampersand (&) and the vertical bar ($|\cdot|$).
- 4. You can filter rows by negating conditions using the exclamation mark (!) logical operator.

5. You can filter rows with missing values using the is.na() function.

The Yaounde COVID-19 dataset

In this lesson, we will again use the data from the COVID-19 serological survey conducted in Yaounde, Cameroon.

```
## # A tibble: 5 × 10
    age sex weight kg highest education neighborhood
  <dbl> <chr> <dbl> <chr> <dbl> <chr>
     45 Female
## 1
                   95 Secondary
                                      Briqueterie
                    96 University
     55 Male
                                      Briqueterie
## 2
                                      Briqueterie
## 3 23 Male
                    74 University
                    70 Secondary
## 4
     20 Female
                                      Briqueterie
## 5 55 Female 67 Primary
                                      Briqueterie
## # ... with 5 more variables: occupation <chr>, is smoker <chr>,
## # is pregnant <chr>, igg result <chr>, igm result <chr>
```

Introducing filter()

We use filter() to keep rows that satisfy a set of conditions. Let's take a look at a simple example. If we want to keep just the male records, we run:

```
yao %>% filter(sex == "Male")
```

```
## # A tibble: 5 × 10
##
   age sex weight kg highest education neighborhood occupation
   <dbl> <chr> <dbl> <chr>
##
                                       <chr>
## 1 55 Male
                  96 University
                                      Briqueterie Salaried worker
                    74 University
## 2
     23 Male
                                      Briqueterie Student
## 3
     28 Male
                   62 Doctorate
                                      Briqueterie Student
## 4 30 Male
## 5 42 Male
                   73 Secondary
                                      Briqueterie Trader
                  71 Secondary
                                       Briqueterie Trader
```

```
## # ... with 4 more variables: is_smoker <chr>, is_pregnant <chr>,
## # igg result <chr>, igm result <chr>
```

Note the use of the double equal sign == rather than the single equal sign =. The == sign tests for equality, as demonstrated below:

```
## create the object `sex_vector` with three elements
sex_vector <- c("Male", "Female",
## test which elements are equal to "Male"
sex_vector == "Male"</pre>
```

```
## [1] TRUE FALSE FALSE
```

So the code yao %>% filter(sex == "Male") will keep all rows where the equality test sex == "Male" evaluates to TRUE.

It is often useful to chain filter() with nrow() to get the number of rows fulfilling a condition.

```
## how many respondents were male?
yao %>%
filter(sex == "Male") %>%
nrow()
```

[1] 422

KEY POINT



The double equal sign, ==, tests for equality, while the single equals sign, =, is used for specifying values to arguments inside functions.



Filter the yao data frame to respondents who were pregnant during the survey.

How many respondents were female? (Use filter() and nrow())

Relational operators

The == operator introduced above is an example of a "relational" operator, as it tests the relation between two values. Here is a list of some of these operators:

Operator	is TRUE if
A < B	A is less than B
A <= B	A is less than or equal to B
A > B	A is greater than B
A >= B	${\sf A}$ is greater than or equal to ${\sf B}$
A == B	A is equal to B
A != B	A is not equal to B
A %in% B	A is an element of B

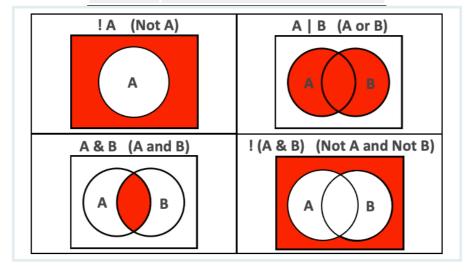


Fig: AND and OR operators visualized.

Let's see how to use these within filter():

```
yao %>% filter(sex != "Male") ## keep rows where `sex` is not "Male"
```

```
## # A tibble: 5 × 10
    age sex weight_kg highest_education neighborhood
##
                <dbl> <chr>
##
   <dbl> <chr>
                                           <chr>
## 1 45 Female
                     95 Secondary
                                          Briqueterie
## 2
      20 Female
                      70 Secondary
                                          Briqueterie
## 3
                      67 Primary
                                          Briqueterie
      55 Female
## 4 17 Female 65 Secondary
## 5 13 Female 65 Secondary
                      65 Secondary
                                           Briqueterie
                                           Briqueterie
## # ... with 5 more variables: occupation <chr>, is smoker <chr>,
## # is_pregnant <chr>, igg_result <chr>, igm_result <chr>
```

yao %>% filter(age < 6) ## keep respondents under 6

```
## # A tibble: 5 × 10
  age sex weight_kg highest_education neighborhood occupation
##
              <dbl> <chr>
                                     <chr> <chr>
   <dbl> <chr>
                 19 Primary
                                                Student
## 1 5 Female
                                     Carriere
## 2
      5 Female
                   26 Primary
                                     Carriere
                                               No response
                   16 Primary
## 3
      5 Male
                                     Cité Verte Student
      5 Female
                   21 Primary
                                     Ekoudou
## 4
                                                Student
## 5 5 Male
                   15 Primary
                                     Ekoudou
                                                Student
## # ... with 4 more variables: is smoker <chr>, is pregnant <chr>,
## # igg result <chr>, igm result <chr>
```

yao %>% filter(age >= 70) ## keep respondents aged at least 70

```
## # A tibble: 5 × 10
##
   age sex weight_kg highest_education neighborhood
95 Secondary
## 1 78 Male
                                  Briqueterie
     79 Female
                  40 Primary
                                  Briqueterie
## 3 78 Female
                  60 Primary
                                  Briqueterie
   75 Male
                                  Briqueterie
                  74 Primary
## 4
## 5 72 Male 65 Secondary
                                  Carriere
## # ... with 5 more variables: occupation <chr>, is smoker <chr>,
## # is pregnant <chr>, igg result <chr>, igm result <chr>
```

```
## keep respondents whose highest education is "Primary" or "Secondary"
yao %>% filter(highest_education %in% c("Primary", "Secondary"))
```

```
## # A tibble: 5 \times 10
   age sex weight_kg highest_education neighborhood
  <dbl> <chr> <dbl> <chr> <dbl> <chr>
                  95 Secondary
## 1 45 Female
                                      Briqueterie
                    70 Secondary
     20 Female
                                      Briqueterie
## 2
     55 Female
## 3
                    67 Primary
                                       Briqueterie
                                      Briqueterie
## 4
     17 Female
                    65 Secondary
## 5 13 Female 65 Secondary Briqueterie
## # ... with 5 more variables: occupation <chr>, is_smoker <chr>,
## # is pregnant <chr>, igg result <chr>, igm result <chr>
```



PRACTICE From yao, keep only respondents who were children (under 18).

With %in%, keep only respondents who live in the "Tsinga" or "Messa" neighborhoods.

Combining conditions with & and |

We can pass multiple conditions to a single filter() statement separated by commas:

```
## keep respondents who are pregnant and are ex-smokers
yao %>% filter(is_pregnant == "Yes", is_smoker == "Ex-smoker") ## only one row
```

When multiple conditions are separated by a comma, they are implicitly combined with an **and** (&).

It is best to replace the comma with & to make this more explicit.

```
## same result as before, but `&` is more explicit
yao %>% filter(is_pregnant == "Yes" & is_smoker == "Ex-smoker")
```

Don't confuse:



- the "," in listing several conditions in filter filter (A,B) i.e. filter based on condition A and (&) condition B
- the "," in lists c(A,B) which is listing different components of the list (and has nothing to do with the & operator)

If we want to combine conditions with an **or**, we use the vertical bar symbol, |.

```
## respondents who are pregnant OR who are ex-smokers
yao %>% filter(is_pregnant == "Yes" | is_smoker == "Ex-smoker")
```



PRACTICE Filter yao to only keep men who tested IgG positive.

Filter yao to keep both children (under 18) and anyone whose highest education is primary school.

Negating conditions with!

To negate conditions, we wrap them in ! ().

Below, we drop respondents who are children (less than 18 years) or who weigh less than 30kg:

```
## drop respondents < 18 years OR < 30 kg
yao %>% filter(!(age < 18 | weight_kg < 30))</pre>
```

```
## # A tibble: 5 × 10
    age sex weight kg highest education neighborhood
## 1 45 Female
                  95 Secondary
                                       Briqueterie
     55 Male
                     96 University
                                       Briqueterie
## 2
                     74 University
                                       Briqueterie
## 3
     23 Male
## 4 20 Female 70 Secondary
## 5 55 Female 67 Primary
                                       Briqueterie
                                       Briqueterie
## # ... with 5 more variables: occupation <chr>, is_smoker <chr>,
## # is pregnant <chr>, igg result <chr>, igm result <chr>
```

The ! operator is also used to negate %in% since R does not have an operator for **NOT in**.

```
## drop respondents whose highest education is NOT "Primary" or "Secondary"
yao %>% filter(!(highest_education %in% c("Primary", "Secondary")))
```

```
## # A tibble: 5 × 10
##
      age sex weight kg highest education neighborhood occupation
   <dbl> <chr> <dbl> <chr> <chr>
##
                                                   <chr>
                   96 University
                                      Briqueterie Salaried worker
## 1 55 Male
## 2
     23 Male
                    74 University
                                      Briqueterie Student
                                       Briqueterie Student
## 3
     28 Male
                    62 Doctorate
## 4
      38 Male
                     71 University
                                       Briqueterie Informal worker
## 5
                   71 University
                                      Briqueterie Salaried worker
     54 Male
## # ... with 4 more variables: is smoker <chr>, is pregnant <chr>,
## # igg result <chr>, igm result <chr>
```

It is easier to read filter() statements as **keep** statements, to avoid confusion over whether we are filtering **in** or filtering **out**!

So the code below would read: "**keep** respondents who are under 18 or who weigh less than 30kg".

KEY POINT



```
yao %>% filter(age < 18 | weight_kg < 30)
```

And when we wrap conditions in ! (), we can then read filter () statements as **drop** statements.

So the code below would read: "**drop** respondents who are under 18 or who weigh less than 30kg".

```
yao %>% filter(!(age < 18 | weight_kg < 30))</pre>
```



From yao, drop respondents who live in the Tsinga or Messa neighborhoods.

NA values

The relational operators introduced so far do not work with NA.

Let's make a data subset to illustrate this.

```
yao_mini <- yao %>%
  select(sex, is_pregnant) %>%
  slice(1,11,50,2) ## custom row order
yao_mini
```

```
## # A tibble: 4 × 2
## sex is_pregnant
## <chr> <chr> <hr > chr> <chr>
## 1 Female No
## 2 Female No response
## 3 Female Yes
## 4 Male <NA>
```

In yao_mini, the last respondent has an NA for the is_pregnant column, because he is male.

Trying to select this row using == NA will not work.

```
yao_mini %>% filter(is_pregnant == NA) ## does not work

## # A tibble: 0 × 2
## # ... with 2 variables: sex <chr>, is_pregnant <chr>

yao_mini %>% filter(is_pregnant == "NA") ## does not work

## # A tibble: 0 × 2
## # ... with 2 variables: sex <chr>, is pregnant <chr>
```

This is because NA is a non-existent value. So R cannot evaluate whether it is "equal to" or "not equal to" anything.

The special function is.na() is therefore necessary:

```
## keep rows where `is_pregnant` is NA
yao_mini %>% filter(is.na(is_pregnant))
```

```
## # A tibble: 1 × 2
## sex is_pregnant
## <chr> <chr>
## 1 Male <NA>
```

This function can be negated with !:

```
## drop rows where `is_pregnant` is NA
yao_mini %>% filter(!is.na(is_pregnant))
```

```
## # A tibble: 3 × 2
## sex is_pregnant
## <chr> <chr>
## 1 Female No
## 2 Female No response
## 3 Female Yes
```

For tibbles, RStudio will highlight ${\tt NA}$ values bright red to distinguish them from other values:



A common error with NA



NA values can be identified but any other encoding such as "NA" or "NaN", which are encoded as strings, will be imperceptible to the functions (they are strings, like any others).



From the yao dataset, keep all the respondents who had missing records for the report of their smoking status.



For some respondents the respiration rate, in breaths per minute, was recorded in the respiration_frequency column.



From yaounde, drop those with a respiration frequency under 20. Think about NAs while doing this! You should avoid also dropping the NA values.

Wrap Up!

Now you know the two essential verbs to select() columns and to filter() rows. This way you keep the variables you are interested in by selecting your columns and you keep the data entries you judge relevant by filtering your rows.

But what about modifying, transforming your data? We will learn about this in the next lesson. See you there!

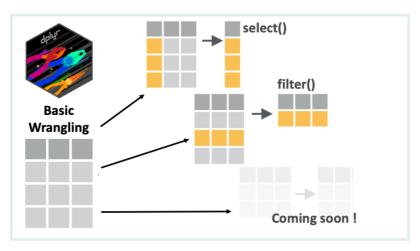


Fig: Basic Data Wrangling: select() and filter().

Contributors

The following team members contributed to this lesson:



LAURE VANCAUWENBERGHE

Data analyst, the GRAPH Network A firm believer in science for good, striving to ally programming, health and education



ANDREE VALLE CAMPOS

R Developer and Instructor, the GRAPH Network

Motivated by reproducible science and education



KENE DAVID NWOSU

Data analyst, the GRAPH Network Passionate about education

References

Some material in this lesson was adapted from the following sources:

- Horst, A. (2021). Dplyr-learnr. https://github.com/allisonhorst/dplyr-learnr (Original work published 2020)
- Subset rows using column values—Filter. (n.d.). Retrieved 12 January 2022, from https://dplyr.tidyverse.org/reference/filter.html

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