

# Why you should use functions

INTRODUCTION TO WRITING FUNCTIONS IN R



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# The arguments to mean()

Mean has 3 arguments

- `x` : A numeric or date-time vector.
- `trim` : The proportion of outliers from each end to remove before calculating
- `na.rm` : Remove before calculating

# Calling mean()

Pass arguments by position

```
mean(numbers, 0.1, TRUE)
```

Pass arguments by name

```
mean(na.rm = TRUE, trim = 0.1, x = numbers)
```

Common arguments by position, rare arguments by name

```
mean(numbers, trim = 0.1, na.rm = TRUE)
```

# Analyzing test scores

```
library(readr)
test_scores_geography_raw <- read_csv("test_scores_geography.csv")
```

```
library(dplyr)
test_scores_geography_clean <- test_scores_geography_raw %>%
  select(person_id, first_name, last_name, test_date, score)
```

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test_scores_geography_clean <- test_scores_geography_raw %>%
  select(person_id, first_name, last_name, test_date, score)
```

```
library(readr)
test_scores_english_raw <- read_csv("test_scores_english.csv")

library(dplyr)
test_scores_english_clean <- test_scores_english_raw %>%
  select(person_id, first_name, last_name, test_date, score)
```

```
library(readr)
test_scores_art_raw <- read_csv("test_scores_art.csv")

library(dplyr)
test_scores_art_clean <- test_scores_art_raw %>%
  select(person_id, first_name, last_name, test_date, score)
```

```
library(readr)
test_scores_spanish_raw <- read_csv("test_scores_spanish.csv")

library(dplyr)
test_scores_spanish_clean <- test_scores_spanish_raw %>%
  select(person_id, first_name, last_name, test_date, score)
```

```
library(readr)
test_scores_geography_raw <- read_csv("test_scores_geography.csv")

library(dplyr)
library(lubridate)
test_scores_geography_clean <- test_scores_geography_raw %>%
  select(person_id, first_name, last_name, test_date, score) %>%
  mutate(test_date = mdy(test_date))
```

```
library(readr)
test_scores_english_raw <- read_csv("test_scores_english.csv")

library(dplyr)
library(lubridate)
test_scores_english_clean <- test_scores_english_raw %>%
  select(person_id, first_name, last_name, test_date, score) %>%
  mutate(test_date = mdy(test_date))
```

```
library(readr)
test_scores_art_raw <- read_csv("test_scores_art.csv")

library(dplyr)
library(lubridate)
test_scores_art_clean <- test_scores_art_raw %>%
  select(person_id, first_name, last_name, test_date, score) %>%
  mutate(test_date = mdy(test_date))
```

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library(readr)
test_scores_spanish_raw <- read_csv("test_scores_spanish.csv")

library(dplyr)
library(lubridate)
test_scores_spanish_clean <- test_scores_spanish_raw %>%
  select(person_id, first_name, last_name, test_date, score) %>%
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  filter(!is.na(score))
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  filter(!is.na(score))
```

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library(readr)
test_scores_art_raw <- read_csv("test_scores_art.csv")

library(dplyr)
library(lubridate)
test_scores_art_clean <- test_scores_art_raw %>%
  select(person_id, first_name, last_name, test_date, score) %>%
  mutate(test_date = mdy(test_date)) %>%
  filter(is.na(score))
```

```
library(readr)
test_scores_spanish_raw <- read_csv("test_scores_spanish.csv")

library(dplyr)
library(lubridate)
test_scores_spanish_clean <- test_scores_spanish_raw %>%
  select(person_id, first_name, last_name, test_date, score) %>%
  mutate(test_date = mdy(test_date)) %>%
  filter(!is.na(score))
```



# Benefits of writing functions

Functions eliminate repetition from your code, which

- can reduce your workload, and
- help avoid errors.

Functions also allow code reuse and sharing.

# Let's practice!

INTRODUCTION TO WRITING FUNCTIONS IN R

# Converting scripts into functions

INTRODUCTION TO WRITING FUNCTIONS IN R



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# A basic function template

```
my_fun <- function(arg1, arg2) {  
  # Do something  
}
```

The signature

```
function(arg1, arg2)
```

The body

```
  {  
  
  # Do something  
}
```

```
library(readr)
test_scores_geography_raw <- read_csv("test_scores_geography.csv")

library(dplyr)
library(lubridate)
test_scores_geography_clean <- test_scores_geography_raw %>%
  select(person_id, first_name, last_name, test_date, score) %>%
  mutate(test_date = mdy(test_date)) %>%
  filter(!is.na(score))
```

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library(readr)
test_scores_english_raw <- read_csv("test_scores_english.csv")

library(dplyr)
library(lubridate)
test_scores_english_clean <- test_scores_english_raw %>%
  select(person_id, first_name, last_name, test_date, score) %>%
  mutate(test_date = mdy(test_date)) %>%
  filter(!is.na(score))
```

```
library(readr)
test_scores_art_raw <- read_csv("test_scores_art.csv")

library(dplyr)
library(lubridate)
test_scores_art_clean <- test_scores_art_raw %>%
  select(person_id, first_name, last_name, test_date, score) %>%
  mutate(test_date = mdy(test_date)) %>%
  filter(is.na(score))
```

```
library(readr)
test_scores_spanish_raw <- read_csv("test_scores_spanish.csv")

library(dplyr)
library(lubridate)
test_scores_spanish_clean <- test_scores_spanish_raw %>%
  select(person_id, first_name, last_name, test_date, score) %>%
  mutate(test_date = mdy(test_date)) %>%
  filter(!is.na(score))
```

# Make a template

```
import_test_scores <- function() {
```

}

# Paste your script into the body

```
import_test_scores <- function() {  
  test_scores_geography_raw <- read_csv("test_scores_geography.csv")  
  
  test_scores_geography_clean <- test_scores_geography_raw %>%  
    select(person_id, first_name, last_name, test_date, score) %>%  
    mutate(test_date = mdy(test_date)) %>%  
    filter(!is.na(score))  
}
```

# Choose the arguments

```
import_test_scores <- function(filename) {           # <- only 1 argument
  test_scores_geography_raw <- read_csv("test_scores_geography.csv")

  test_scores_geography_clean <- test_scores_geography_raw %>%
    select(person_id, first_name, last_name, test_date, score) %>%
    mutate(test_date = mdy(test_date)) %>%
    filter(!is.na(score))
}
```



# Replace specific values with arguments

```
import_test_scores <- function(filename) {  
  test_scores_geography_raw <- read_csv(filename) # <- replace specific filename  
  
  test_scores_geography_clean <- raw_data %>%  
    select(person_id, first_name, last_name, test_date, score) %>%  
    mutate(test_date = mdy(test_date)) %>%  
    filter(!is.na(score))  
}
```

# Generalize variable names

```
import_test_scores <- function(filename) {  
  test_scores_raw <- read_csv(filename)  # <- variable names generalized  
  
  test_scores_clean <- test_scores_raw %>%      # <- variable names generalized  
    select(person_id, first_name, last_name, test_date, score) %>%  
    mutate(test_date = mdy(test_date)) %>%  
    filter(!is.na(score))  
}
```

# Remove the final assignment

```
import_test_scores <- function(filename) {  
  test_scores_raw <- read_csv(filename)  
  
  test_scores_raw %>% # <- remove assignment  
    select(person_id, first_name, last_name, test_date, score) %>%  
    mutate(test_date = mdy(test_date)) %>%  
    filter(!is.na(score))  
}
```

# Use your function

```
test_scores_geography <- import_test_scores("test_scores_geography.csv")
test_scores_english <- import_test_scores("test_scores_english.csv")
test_scores_art <- import_test_scores("test_scores_art.csv")
test_scores_spanish <- import_test_scores("test_scores_spanish.csv")
```

# Arguments of sample()

- `x` : A vector of values to sample from.
- `size` : How many times do you want to sample from `x` ?
- `replace` : Should you sample with replacement or not?
- `prob` : A vector of sampling weights for each value of `x` , totaling one.

# Let's practice!

INTRODUCTION TO WRITING FUNCTIONS IN R

# Y kant I reed ur code?

INTRODUCTION TO WRITING FUNCTIONS IN R



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# dplyr verbs

`select()` *selects* columns

`filter()` *filters* rows



# Function names should contain a verb

- get
- calculate (or maybe just calc)
- run
- process
- import
- clean
- tidy
- draw

# lm() is badly named

- Acronyms aren't self-explanatory
- It doesn't contain a verb
- There are lots of different linear models

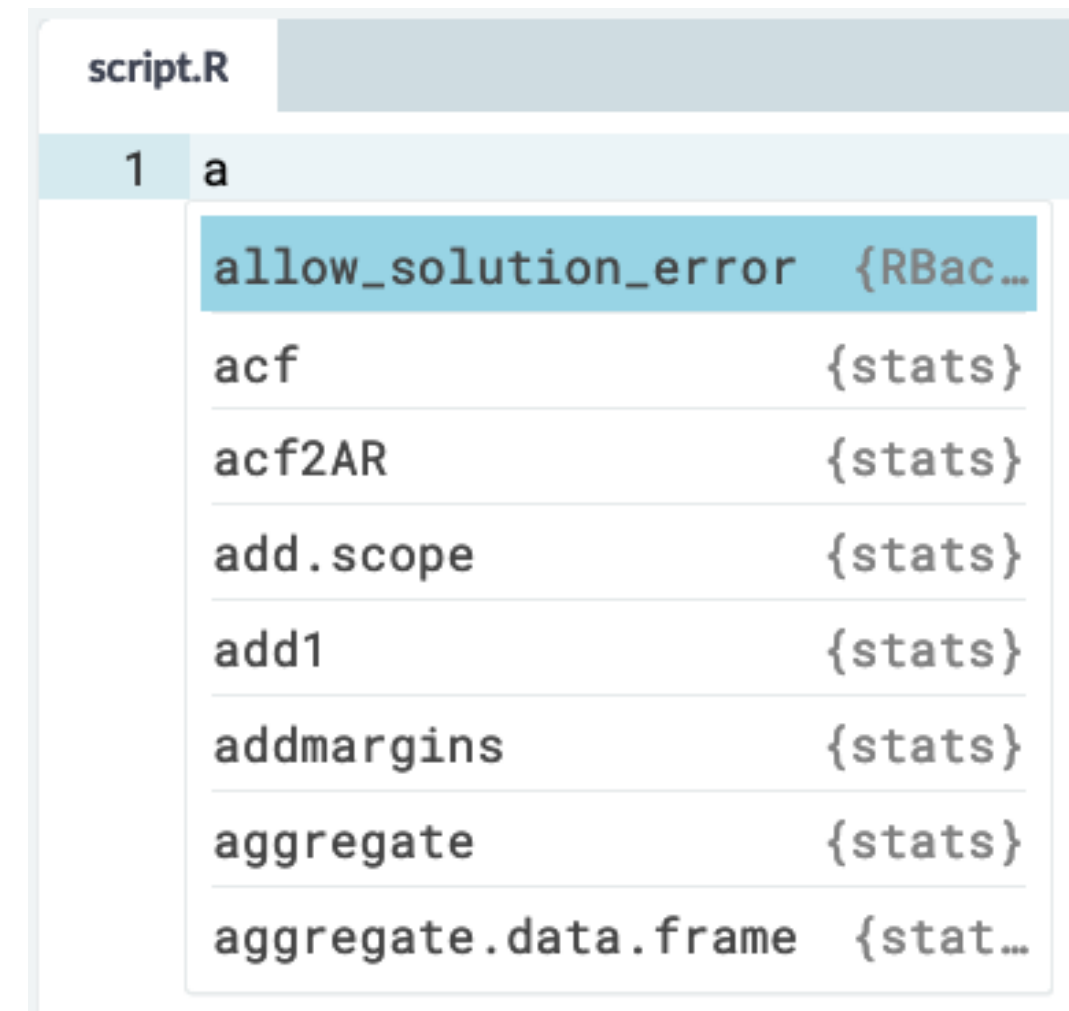
A better name would be `run_linear_regression()`

# Readability vs. typeability

- Understanding code >> typing code

# Readability vs. typeability

- Understanding code >> typing code
- Code editors have autocomplete



The screenshot shows a code editor window titled 'script.R'. On line 1, the character 'a' is typed. A dropdown menu is open, displaying a list of R functions that start with 'a'. The first option, 'allow\_solution\_error {RBac...', is highlighted in blue. The other options are 'acf {stats}', 'acf2AR {stats}', 'add.scope {stats}', 'add1 {stats}', 'addmargins {stats}', 'aggregate {stats}', and 'aggregate.data.frame {stat...'.

Function	Package
allow_solution_error	{RBac...
acf	{stats}
acf2AR	{stats}
add.scope	{stats}
add1	{stats}
addmargins	{stats}
aggregate	{stats}
aggregate.data.frame	{stat...

# Readability vs. typeability

- Understanding code >> typing code
- Code editors have autocomplete
- You can alias common functions

```
h <- head
```

```
data(cats, package = "MASS")  
h(cats)
```

```
  Sex Bwt Hwt  
1   F 2.0 7.0  
2   F 2.0 7.4  
3   F 2.0 9.5  
4   F 2.1 7.2  
5   F 2.1 7.3  
6   F 2.1 7.6
```

# Arguments of lm()

args(lm)

```
function (formula, data, subset, weights, na.action, method = "qr",  
         model = TRUE, x = FALSE, y = FALSE, qr = TRUE, singular.ok = TRUE,  
         contrasts = NULL, offset, ...)
```

# Types of argument

- **Data arguments:** what you compute on
- **Detail arguments:** how you perform the computation

```
args(cor)
```

```
function (x, y = NULL, use = "everything",  
          method = c("pearson", "kendall", "spearman"))
```

# Data args should precede detail args

This won't work

```
data %>%  
  lm(formula)
```

because the data argument isn't first.



# Our revised function for linear regression

```
run_linear_regression <- function(data, formula) {  
  lm(formula, data)  
}
```

```
cats %>%  
  run_linear_regression(Hwt ~ Bwt + Sex)
```

```
Call:  
lm(formula = formula, data = data)  
  
Coefficients:  
(Intercept)          Bwt          SexM  
    -0.4150      4.0758     -0.0821
```

# Let's practice!

INTRODUCTION TO WRITING FUNCTIONS IN R