

# R and the tidyverse

## Winter Institute in Data Science

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2025-12-15

What is R?

R Functions

Data Structures

Core tidyverse Transformation Functions

Other Common Transformation Functions

Helper Functions

## R + RTM

- ▶ ≈ 23rd anniversary

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  - ▶ Web-scraping, mapping, mail-merging (`muRL`)

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  - ▶ Functions for displays, geolocated data, etc. (`labtoolbox`)
- ▶ Research in R, teach with R, teach R, consult with R, run family gift exchange with R, War,  
...

What is R?

# What is R?

“R is a language and environment  
for statistical computing and graphics”

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- ▶ Software for calculation, computation, data analysis

## What is R?

“R is a language and environment  
for statistical computing and graphics”

- ▶ Software for calculation, computation, data analysis
- ▶ Well-developed graphical facilities
- ▶ A programming language

# Why use R (and Python)?

- ▶ Standard for data analysis, modeling, graphics
- ▶ High-quality, powerful, flexible, extensible
- ▶ International community (including here)
- ▶ Platform independent (Mac OS, Windows, Linux/Unix)
- ▶ Free + Open Source
- ▶ Reads `.xlsx`, `.dta`, `.csv`, `.txt`, `.json`, ...
- ▶ Interfaces with C, C++, Ruby, Java, R/Python, Unix, ...
- ▶ Command line (Mac OS Terminal prompt),  
Windows/Mac/Linux GUIs
- ▶ RStudio: excellent IDE (code, plots, etc. 1 window;  
GitHub)
- ▶ Let R teach you R: `swirl`

# How does R Work?

# How does R Work?

5 + 2

# How does R Work?

5 + 2

```
## [1] 7
```

## How does R Work?

```
5 + 2
```

```
## [1] 7
```

```
sum(5, 2)
```

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```
5 + 2
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sum(5, 2)
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## [1] 7
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But not just printing:

# How does R Work?

```
5 + 2
```

```
## [1] 7
```

```
sum(5, 2)
```

```
## [1] 7
```

But not just printing:

```
a <- sum(5, 2)
b <- median(1:10)
a + b # (Hi -- Notes after the '#' R ignores)
```

## How does R Work?

```
5 + 2
```

```
## [1] 7
```

```
sum(5, 2)
```

```
## [1] 7
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But not just printing:

```
a <- sum(5, 2)
b <- median(1:10)
a + b # (Hi -- Notes after the '#' R ignores)
```

```
## [1] 12.5
```

# How does R Work?

```
5 + 2
```

```
## [1] 7
```

```
sum(5, 2)
```

```
## [1] 7
```

But not just printing:

```
a <- sum(5, 2)
b <- median(1:10)
a + b # (Hi -- Notes after the '#' R ignores)
```

```
## [1] 12.5
```

```
difftime("2026-07-04", "2025-12-15")
```

# How does R Work?

```
5 + 2
```

```
## [1] 7
```

```
sum(5, 2)
```

```
## [1] 7
```

But not just printing:

```
a <- sum(5, 2)  
b <- median(1:10)
```

```
a + b # (Hi -- Notes after the '#' R ignores)
```

```
## [1] 12.5
```

```
difftime("2026-07-04", "2025-12-15")
```

```
## Time difference of 200.9583 days
```

# How to Work in R

- ▶ Open R/RStudio

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- ▶ Open R/RStudio
- ▶ Create a .R file

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  - ~~ Mac: Cmd-Return to execute a line  
(better than copy-paste)
  - ~~ At >, [Up Arrow] recalls previous command

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- ▶ Quit

# How to Work in R

- ▶ Open R/RStudio
- ▶ Create a .R file
- ▶ Add code and comments to the .R file
- ▶ Run them to get output, results, graphics, ...
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(better than copy-paste)
  - ~~ At >, [Up Arrow] recalls previous command
- ▶ Save .R file
- ▶ Quit (do not save workspace)

# How to Work in R

- ▶ Open R/RStudio
- ▶ Create a .R file
- ▶ Add code and comments to the .R file
- ▶ Run them to get output, results, graphics, ...
  - ~~ Mac: Cmd-Return to execute a line  
(better than copy-paste)
  - ~~ At >, [Up Arrow] recalls previous command
- ▶ Save .R file
- ▶ Quit (do not save workspace)

Later, ...

- ▶ Open .R file
- ▶ Add more code and comments ...

# How to Work in R

## *Literate programming*

- ▶ In addition to `.R`, we use `.qmd` (Quarto) and `.Rmd` (RMarkdown)

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## *Literate programming*

- ▶ In addition to `.R`, we use `.qmd` (Quarto) and `.Rmd` (RMarkdown)
- ▶ Interweave code with text/images, output ↠ PDF/HTML

# How to Work in R

## *Literate programming*

- ▶ In addition to `.R`, we use `.qmd` (Quarto) and `.Rmd` (RMarkdown)
- ▶ Interweave code with text/images, output ↠ PDF/HTML
- ▶ (More later today...)

# How do I get help?

Within R:

```
help(mean)  
help.search("median")
```

## How do I get help?

Within R:

```
help(mean)  
help.search("median")
```

```
example(mean)
```

```
##  
## mean> x <- c(0:10, 50)  
##  
## mean> xm <- mean(x)  
##  
## mean> c(xm, mean(x, trim = 0.10))  
## [1] 8.75 5.50
```

# How do I get help?

Outside of R:

- ▶ Q & A
  - ▶ Stack Overflow (tags `r`, `rstats`)

# How do I get help?

Outside of R:

- ▶ Q & A
  - ▶ Stack Overflow (tags `r`, `rstats`)
- ▶ Courses and references
  - ▶ `rseek.org` (custom Google search)
  - ▶ `CRANsearcher` (RStudio add-in for pkgs)
  - ▶ Data Camp/Lynda.com video courses through AU Portal
  - ▶ Many good books and documents: Cookbook, Intro Statistics, Student Companion, Graphics, Mapping, Programming, Short Ref Card, ...
  - ▶ Wickham, Çetinkaya-Rundel, and Grolemund (2023)

## R Functions

## Functions

```
function(arg1, arg2, ...){  
    <the function code here...>  
}
```

## Functions

```
function(arg1, arg2, ...){  
  <the function code here...>  
}
```

```
sum(5, 2)
```

```
## [1] 7
```

## Functions

```
function(arg1, arg2, ...){  
  <the function code here...>  
}
```

```
sum(5, 2)
```

```
## [1] 7
```

```
mean(1:4)
```

```
## [1] 2.5
```

## Functions

```
## [1] "(Ready, Kurt?)"
```

## Functions

```
## [1] "(Ready, Kurt?)"
```

```
nchar("greetings")
```

## Functions

```
## [1] "(Ready, Kurt?)"
```

```
nchar("greetings")
```

```
## [1] 9
```

## Functions

```
## [1] "(Ready, Kurt?)"
```

```
nchar("greetings")
```

```
## [1] 9
```

```
## [1] "(Ready, Will?)"
```

## Functions

```
## [1] "(Ready, Kurt?)"
```

```
nchar("greetings")
```

```
## [1] 9
```

```
## [1] "(Ready, Will?)"
```

```
length(us)
```

## Functions

```
## [1] "(Ready, Kurt?)"
```

```
nchar("greetings")
```

```
## [1] 9
```

```
## [1] "(Ready, Will?)"
```

```
length(us)
```

```
## [1] 10
```

## A Useful Function: `c()`

To concatenate objects into a vector, use `c()`:

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```
c(1, 3, 8, 20)
```

```
## [1] 1 3 8 20
```

## A Useful Function: `c()`

To concatenate objects into a vector, use `c()`:

```
c(1, 3, 8, 20)
```

```
## [1] 1 3 8 20
```

```
c("a", "merican", "u")
```

## A Useful Function: `c()`

To concatenate objects into a vector, use `c()`:

```
c(1, 3, 8, 20)
```

```
## [1] 1 3 8 20
```

```
c("a", "merican", "u")
```

```
## [1] "a"          "merican"    "u"
```

## A Useful Function: `c()`

To concatenate objects into a vector, use `c()`:

```
c(1, 3, 8, 20)
```

```
## [1] 1 3 8 20
```

```
c("a", "merican", "u")
```

```
## [1] "a"          "merican"    "u"
```

```
c(1, 2, "hello")
```

## A Useful Function: `c()`

To concatenate objects into a vector, use `c()`:

```
c(1, 3, 8, 20)
```

```
## [1] 1 3 8 20
```

```
c("a", "merican", "u")
```

```
## [1] "a"         "merican"   "u"
```

```
c(1, 2, "hello")
```

```
## [1] "1"        "2"        "hello"
```

## Functions' Arguments

What arguments does a function have?

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```
help(median)
```

```
args(median)
```

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```
help(median)
```

```
args(median)
```

## Functions' Arguments

```
median(1:3)
```

```
## [1] 2
```

## Functions' Arguments

```
median(1:3)
```

```
## [1] 2
```

```
x <- c(1, 2, 3, NA)
median(x)
```

## Functions' Arguments

```
median(1:3)
```

```
## [1] 2
```

```
x <- c(1, 2, 3, NA)
median(x)
```

```
## [1] NA
```

## Functions' Arguments

```
median(1:3)
```

```
## [1] 2
```

```
x <- c(1, 2, 3, NA)
median(x)
```

```
## [1] NA
```

```
median(x, na.rm = TRUE)
```

## Functions' Arguments

```
median(1:3)
```

```
## [1] 2
```

```
x <- c(1, 2, 3, NA)
median(x)
```

```
## [1] NA
```

```
median(x, na.rm = TRUE)
```

```
## [1] 2
```

## Functions' Arguments

You can specify arguments in order or by name:

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You can specify arguments in order or by name:

```
median(x, TRUE)
```

```
## [1] 2
```

## Functions' Arguments

You can specify arguments in order or by name:

```
median(x, TRUE)
```

```
## [1] 2
```

```
median(na.rm = TRUE, x)
```

```
## [1] 2
```

## Functions' Arguments

You can specify arguments in order or by name:

```
median(x, TRUE)
```

```
## [1] 2
```

```
median(na.rm = TRUE, x)
```

```
## [1] 2
```

```
median(TRUE, x)
```

## Functions' Arguments

You can specify arguments in order or by name:

```
median(x, TRUE)
```

```
## [1] 2
```

```
median(na.rm = TRUE, x)
```

```
## [1] 2
```

```
median(TRUE, x)
```

```
## Error in if (na.rm) x <- x[!is.na(x)] else if (any(:
```

## Some Useful Functions

Managing the workspace:

```
# Get the working directory ("Where am I?"):  
getwd()
```

```
## [1] "/Users/rtm/Documents/github/winter-inst/01-intro
```

## Some Useful Functions

Managing the workspace:

```
# Get the working directory ("Where am I?"):
getwd()
```

```
## [1] "/Users/rtm/Documents/github/winter-inst/01-intro
```

```
# Set the working directory:
setwd("~/Desktop/")
```

## Some Useful Functions

Managing the workspace:

```
# Get the working directory ("Where am I?"):  
getwd()
```

```
## [1] "/Users/rtm/Documents/github/winter-inst/01-intro"
```

```
# Set the working directory:  
setwd("~/Desktop/")
```

Better, use “R projects” and the `here` package:

```
library(here)
```

```
## here() starts at /Users/rtm/Documents/github/winter-
```

## Some Useful Functions

Managing the workspace:

```
# Get the working directory ("Where am I?"):  
getwd()
```

```
## [1] "/Users/rtm/Documents/github/winter-inst/01-intro"
```

```
# Set the working directory:  
setwd("~/Desktop/")
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Better, use “R projects” and the `here` package:

```
library(here)
```

```
## here() starts at /Users/rtm/Documents/github/winter-
```

```
here()
```

```
## [1] "/Users/rtm/Documents/github/winter-inst" 74 / 274
```

## Some Useful Functions

Managing the workspace:

```
# List objects in working dir:  
ls()
```

```
## [1] "a"     "b"     "tmp"   "us"    "x"     "xm"
```

```
# Remove 'x' from working dir:  
rm(x)
```

```
# Remove everything from working dir:  
# rm(list = ls())
```

# Some Useful Functions

Managing the workspace:

```
# List objects in working dir:  
ls()
```

```
## [1] "a"     "b"     "tmp"   "us"    "x"     "xm"
```

```
# Remove `x' from working dir:  
rm(x)
```

```
# Remove everything from working dir:  
# rm(list = ls())
```

Better, start a fresh session to *really* reset environment:

RStudio - Session - Restart R

Mac: Shift - Cmd - O

## Some Useful Mathematical Functions

```
5 + 2
```

```
## [1] 7
```

```
5 - 2
```

```
## [1] 3
```

```
5 * 2
```

```
## [1] 10
```

```
5 / 2
```

```
## [1] 2.5
```

## Some Useful Mathematical Functions

```
5^2
```

```
## [1] 25
```

```
sqrt(25)
```

```
## [1] 5
```

```
20 %% 3
```

```
## [1] 2
```

# Some Useful Mathematical Functions and Values

```
pi
```

```
## [1] 3.141593
```

```
abs(-3)
```

```
## [1] 3
```

```
exp(1)
```

```
## [1] 2.718282
```

```
log(exp(2))
```

```
## [1] 2
```

```
sin(pi / 2)
```

```
## [1] 1
```

# Logicals

```
TRUE
```

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

```
FALSE
```

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

# Logicals

```
TRUE
```

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

```
FALSE
```

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

```
TRUE == FALSE
```

# Logicals

```
TRUE
```

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

```
FALSE
```

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

```
TRUE == FALSE
```

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

## Logicals

```
c(1, 2) == c(1, 3)
```

## Logicals

```
c(1, 2) == c(1, 3)
```

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

## Logicals

```
c(1, 2) == c(1, 3)
```

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

```
c(1, 2) != c(1, 3)
```

## Logicals

```
c(1, 2) == c(1, 3)
```

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

```
c(1, 2) != c(1, 3)
```

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

## Logicals

```
c(1, 2) == c(1, 3)
```

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

```
c(1, 2) != c(1, 3)
```

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

```
c(1, 2) < c(1, 3)
```

## Logicals

```
c(1, 2) == c(1, 3)
```

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

```
c(1, 2) != c(1, 3)
```

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

```
c(1, 2) < c(1, 3)
```

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

## Logicals

```
c(1, 2) > c(1, 3)
```

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

```
c(1, 2) <= c(1, 3)
```

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

```
c(1, 2) >= c(1, 3)
```

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

## How to Write a New Function

```
sumDiff <- function(num1 = 3, num2 = 5){  
  sum <- num1 + num2  
  diff <- num1 - num2  
  return(c(sum, diff))  
}
```

## How to Write a New Function

```
sumDiff <- function(num1 = 3, num2 = 5){  
  sum <- num1 + num2  
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  return(c(sum, diff))  
}
```

Now, cut and paste function into R prompt.

## How to Write a New Function

```
sumDiff <- function(num1 = 3, num2 = 5){  
  sum <- num1 + num2  
  diff <- num1 - num2  
  return(c(sum, diff))  
}
```

Now, cut and paste function into R prompt.  
(R will tell you if syntax error.)

# My New Function

**sumDiff()**

## My New Function

```
sumDiff()
```

```
## [1] 8 -2
```

## My New Function

```
sumDiff()
```

```
## [1] 8 -2
```

```
sumDiff(3, 5)
```

## My New Function

```
sumDiff()
```

```
## [1] 8 -2
```

```
sumDiff(3, 5)
```

```
## [1] 8 -2
```

## My New Function

```
sumDiff()
```

```
## [1] 8 -2
```

```
sumDiff(3, 5)
```

```
## [1] 8 -2
```

```
sumDiff(num2 = 5, num1 = 3)
```

## My New Function

```
sumDiff()
```

```
## [1] 8 -2
```

```
sumDiff(3, 5)
```

```
## [1] 8 -2
```

```
sumDiff(num2 = 5, num1 = 3)
```

```
## [1] 8 -2
```

## My New Function

```
sumDiff()
```

```
## [1] 8 -2
```

```
sumDiff(3, 5)
```

```
## [1] 8 -2
```

```
sumDiff(num2 = 5, num1 = 3)
```

```
## [1] 8 -2
```

```
sumDiff(5, 3)
```

## My New Function

```
sumDiff()
```

```
## [1] 8 -2
```

```
sumDiff(3, 5)
```

```
## [1] 8 -2
```

```
sumDiff(num2 = 5, num1 = 3)
```

```
## [1] 8 -2
```

```
sumDiff(5, 3)
```

```
## [1] 8 2
```

## My New Function

```
sumDiff(2, 20)
```

## My New Function

```
sumDiff(2, 20)
```

```
## [1] 22 -18
```

## My New Function

```
sumDiff(2, 20)
```

```
## [1] 22 -18
```

```
sumDiff(1, "yes")
```

## My New Function

```
sumDiff(2, 20)
```

```
## [1] 22 -18
```

```
sumDiff(1, "yes")
```

```
## Error in num1 + num2: non-numeric argument
```

# Data Structures

# Data Types

- ▶ Numeric
- ▶ Integer
- ▶ Complex
- ▶ Logical
- ▶ Character
- ▶ Factor

# Data Types

- ▶ Numeric
- ▶ Integer
- ▶ Complex
- ▶ Logical
- ▶ Character
- ▶ Factor
  - ~~ categorical vars: stored as numeric, but w/ char label
  - ~~ great for statistical modeling, graphics (auto indicators, e.g.)

# Data Structures

- ▶ Scalar
- ▶ Vector
- ▶ Matrix

# Data Structures

- ▶ Scalar
- ▶ Vector
- ▶ Matrix
- ▶ Data frame (like matrix, w/ attributes)

# Data Structures

- ▶ Scalar
- ▶ Vector
- ▶ Matrix
- ▶ Data frame (like matrix, w/ attributes)
- ▶ Tibble (tidyverse dataframe)

# Data Structures

- ▶ Scalar
- ▶ Vector
- ▶ Matrix
- ▶ Data frame (like matrix, w/ attributes)
- ▶ Tibble (tidyverse dataframe)
- ▶ List (flexible storage; regression/cluster output)

What is this thing?

```
x <- 1:4  
is.vector(x)
```

What is this thing?

```
x <- 1:4  
is.vector(x)
```

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

What is this thing?

```
x <- 1:4  
is.vector(x)
```

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

```
is.numeric(x)
```

What is this thing?

```
x <- 1:4
```

```
is.vector(x)
```

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

```
is.numeric(x)
```

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

What is this thing?

```
x <- 1:4
```

```
is.vector(x)
```

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

```
is.numeric(x)
```

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

```
is.character(x)
```

What is this thing?

```
x <- 1:4
```

```
is.vector(x)
```

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

```
is.numeric(x)
```

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

```
is.character(x)
```

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

What is this thing?

```
y <- c("a", "hello")
is.vector(y)
```

What is this thing?

```
y <- c("a", "hello")  
is.vector(y)
```

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

What is this thing?

```
y <- c("a", "hello")  
is.vector(y)
```

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

```
is.numeric(y)
```

What is this thing?

```
y <- c("a", "hello")  
is.vector(y)
```

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

```
is.numeric(y)
```

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

What is this thing?

```
y <- c("a", "hello")  
is.vector(y)
```

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

```
is.numeric(y)
```

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

```
is.character(y)
```

What is this thing?

```
y <- c("a", "hello")  
is.vector(y)
```

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

```
is.numeric(y)
```

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

```
is.character(y)
```

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

What is this thing?

```
z <- c(1, 2, 3, NA)  
isNAz <- is.na(z)
```

What is this thing?

```
z <- c(1, 2, 3, NA)  
isNAz <- is.na(z)
```

```
isNAz
```

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

What is this thing?

```
z <- c(1, 2, 3, NA)  
isNAz <- is.na(z)
```

```
isNAz
```

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

```
sum(isNAz)
```

What is this thing?

```
z <- c(1, 2, 3, NA)  
isNAz <- is.na(z)
```

```
isNAz
```

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

```
sum(isNAz)
```

```
## [1] 1
```

What is this thing?

```
z <- c(1, 2, 3, NA)
```

```
isNAz <- is.na(z)
```

```
isNAz
```

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

```
sum(isNAz)
```

```
## [1] 1
```

```
mean(isNAz)
```

What is this thing?

```
z <- c(1, 2, 3, NA)
```

```
isNAz <- is.na(z)
```

```
isNAz
```

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

```
sum(isNAz)
```

```
## [1] 1
```

```
mean(isNAz)
```

```
## [1] 0.25
```

What is this thing?

```
z <- c(1, 2, 3, NA)
```

```
isNAz <- is.na(z)
```

```
isNAz
```

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

```
sum(isNAz)
```

```
## [1] 1
```

```
mean(isNAz)
```

```
## [1] 0.25
```

(“coercion”)

## Data from Where?

- ▶ From the keyboard
- ▶ From within a package
- ▶ From .RData/.RDS file
- ▶ From a local .txt, .csv, .dta, .xlsx, etc. file
- ▶ From a remote file on the web
- ▶ From remote HTML

## Data: Extracting and Assigning Vector Elements

```
y <- c(20, 20, 30, 70, 10)  
x <- c(10, 20, 30, 40, 25)  
x[2]
```

## Data: Extracting and Assigning Vector Elements

```
y <- c(20, 20, 30, 70, 10)
x <- c(10, 20, 30, 40, 25)
x[2]
```

```
## [1] 20
```

## Data: Extracting and Assigning Vector Elements

```
y <- c(20, 20, 30, 70, 10)
x <- c(10, 20, 30, 40, 25)
x[2]
```

```
## [1] 20
```

```
y[3:5]
```

## Data: Extracting and Assigning Vector Elements

```
y <- c(20, 20, 30, 70, 10)
x <- c(10, 20, 30, 40, 25)
x[2]
```

```
## [1] 20
```

```
y[3:5]
```

```
## [1] 30 70 10
```

## Data: Extracting and Assigning Vector Elements

```
y <- c(20, 20, 30, 70, 10)
x <- c(10, 20, 30, 40, 25)
x[2]
```

```
## [1] 20
```

```
y[3:5]
```

```
## [1] 30 70 10
```

```
x[c(1, 5)]
```

## Data: Extracting and Assigning Vector Elements

```
y <- c(20, 20, 30, 70, 10)
x <- c(10, 20, 30, 40, 25)
x[2]
```

```
## [1] 20
```

```
y[3:5]
```

```
## [1] 30 70 10
```

```
x[c(1, 5)]
```

```
## [1] 10 25
```

## Data: Extracting and Assigning Vector Elements

```
y <- c(20, 20, 30, 70, 10)
x <- c(10, 20, 30, 40, 25)
x[2]
```

```
## [1] 20
```

```
y[3:5]
```

```
## [1] 30 70 10
```

```
x[c(1, 5)]
```

```
## [1] 10 25
```

```
x[3] <- 100
```

```
x
```

## Data: Extracting and Assigning Vector Elements

```
y <- c(20, 20, 30, 70, 10)
x <- c(10, 20, 30, 40, 25)
x[2]
```

```
## [1] 20
```

```
y[3:5]
```

```
## [1] 30 70 10
```

```
x[c(1, 5)]
```

```
## [1] 10 25
```

```
x[3] <- 100
```

```
x
```

```
## [1] 10 20 100 40 25
```

## Data: Extracting and Assigning Matrix Elements

```
m <- matrix(c(20, 20, 30, 10, 20, 30), 3, 2)
```

```
m
```

```
##      [,1] [,2]
## [1,]    20   10
## [2,]    20   20
## [3,]    30   30
```

## Data: Extracting and Assigning Matrix Elements

```
m <- matrix(c(20, 20, 30, 10, 20, 30), 3, 2)
```

```
m
```

```
##      [,1] [,2]
## [1,]    20   10
## [2,]    20   20
## [3,]    30   30
```

```
m[1, 2]
```

## Data: Extracting and Assigning Matrix Elements

```
m <- matrix(c(20, 20, 30, 10, 20, 30), 3, 2)
```

```
m
```

```
##      [,1] [,2]  
## [1,]    20   10  
## [2,]    20   20  
## [3,]    30   30
```

```
m[1, 2]
```

```
## [1] 10
```

## Data: Extracting and Assigning Matrix Elements

```
m <- matrix(c(20, 20, 30, 10, 20, 30), 3, 2)
```

```
m
```

```
##      [,1] [,2]  
## [1,]    20   10  
## [2,]    20   20  
## [3,]    30   30
```

```
m[1, 2]
```

```
## [1] 10
```

```
m[2, 2] <- NA
```

```
m
```

## Data: Extracting and Assigning Matrix Elements

```
m <- matrix(c(20, 20, 30, 10, 20, 30), 3, 2)
```

```
m
```

```
##      [,1] [,2]  
## [1,]    20   10  
## [2,]    20   20  
## [3,]    30   30
```

```
m[1, 2]
```

```
## [1] 10
```

```
m[2, 2] <- NA
```

```
m
```

```
##      [,1] [,2]  
## [1,]    20   10  
## [2,]    20   NA  
## [3,]    30   30
```

## Data from Keyboard, into a Data Frame

```
y <- c(20, 20, 30, 70, 10)
x <- c(10, 20, 30, 40, 25)
df <- data.frame(age = y, score = x)
```

## Data from Keyboard, into a Data Frame

```
y <- c(20, 20, 30, 70, 10)
x <- c(10, 20, 30, 40, 25)
df <- data.frame(age = y, score = x)
```

```
df
```

```
##   age score
## 1   20    10
## 2   20    20
## 3   30    30
## 4   70    40
## 5   10    25
```

## Data from Keyboard, into a Data Frame

```
df$age
```

```
## [1] 20 20 30 70 10
```

## Data from Keyboard, into a Data Frame

```
df$age
```

```
## [1] 20 20 30 70 10
```

```
rownames(df)
```

```
## [1] "1" "2" "3" "4" "5"
```

## Data from Keyboard, into a Data Frame

```
df$age
```

```
## [1] 20 20 30 70 10
```

```
rownames(df)
```

```
## [1] "1" "2" "3" "4" "5"
```

```
colnames(df)
```

```
## [1] "age"    "score"
```

## Data: From Keyboard, into a List

```
my_list <- list(x = 1:3, y = letters[1:5],  
                 final = matrix(1:4, 2, 2))
```

## Data: From Keyboard, into a List

```
my_list <- list(x = 1:3, y = letters[1:5],  
                 final = matrix(1:4, 2, 2))
```

```
my_list
```

```
## $x  
## [1] 1 2 3  
##  
## $y  
## [1] "a" "b" "c" "d" "e"  
##  
## $final  
##      [,1] [,2]  
## [1,]     1     3  
## [2,]     2     4
```

## Data: From Keyboard, into a List

```
my_list[[1]]
```

## Data: From Keyboard, into a List

```
my_list[[1]]
```

```
## [1] 1 2 3
```

## Data: From Keyboard, into a List

```
my_list[[1]]
```

```
## [1] 1 2 3
```

```
my_list[["final"]]
```

## Data: From Keyboard, into a List

```
my_list[[1]]
```

```
## [1] 1 2 3
```

```
my_list[["final"]]
```

```
##          [,1] [,2]
## [1,]      1     3
## [2,]      2     4
```

## Data: Lists

```
my_list$x
```

```
## [1] 1 2 3
```

```
my_list$y
```

```
## [1] "a" "b" "c" "d" "e"
```

## Data: Lists

```
my_list$x
```

```
## [1] 1 2 3
```

```
my_list$y
```

```
## [1] "a" "b" "c" "d" "e"
```

A data frame is a list.

## Data from a Package

```
library(car)  
data(Chile)
```

## Data from a Package

```
library(car)
data(Chile)
```

```
head(Chile)
```

```
##   region population sex age education income status
## 1     N      175000  M  65          P  35000  1.008
## 2     N      175000  M  29          PS   7500 -1.296
## 3     N      175000  F  38          P  15000  1.230
## 4     N      175000  F  49          P  35000 -1.031
## 5     N      175000  F  23          S  35000 -1.104
## 6     N      175000  F  28          P   7500 -1.046
```

## Core tidyverse Transformation Functions

## What is a Package?

An R package is an extension of R that includes

- ▶ a set of functions for users
- ▶ datasets
- ▶ demonstration code
- ▶ “background” code  
(in R or a compiled language)
- ▶ documentation
- ▶ metadata  
(authors, license, e.g.)

How do I get package `thispackage`?

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```
install.packages("thispackage")
```

(Once, per R version, at the Console.)

How do I get package `thispackage`?

```
install.packages("thispackage")
```

(Once, per R version, at the Console.)

Then, in my R file,

```
library(thispackage)
```

When we refer to a package (in English), we sometimes put it inside { }:

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- ▶ “Consider the `{lubridate}` package for that problem.”

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When we explicitly refer to a function from a package (in English or in code), we use ::

When we refer to a package (in English), we sometimes put it inside { }:

- ▶ “Consider the `{lubridate}` package for that problem.”

When we explicitly refer to a function from a package (in English or in code), we use ::

- ▶ English: “Use `janitor::clean_names()`”
- ▶ code:

```
df <- df |> janitor::clean_names()
```

## What is the tidyverse?

Coherent, consistent *set* of R packages for data import, manipulation, visualization, etc.

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Coherent, consistent *set* of R packages for data import, manipulation, visualization, etc.

How do I get the tidyverse?

```
## [1] "(Ready, Rian?)"
```

## What is the tidyverse?

Coherent, consistent *set* of R packages for data import, manipulation, visualization, etc.

How do I get the tidyverse?

```
## [1] "(Ready, Rian?)"
```

## What is the tidyverse?

Coherent, consistent *set* of R packages for data import, manipulation, visualization, etc.

How do I get the tidyverse?

```
## [1] "(Ready, Rian?)"
```

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

## What is the tidyverse?

Coherent, consistent *set* of R packages for data import, manipulation, visualization, etc.

How do I get the tidyverse?

```
## [1] "(Ready, Rian?)"
```

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

```
library(tidyverse)
```

## The Core tidyverse Transformation Functions

First:

```
library(tidyverse)
```

## The Core tidyverse Transformation Functions

First:

```
library(tidyverse)
```

```
(or just library(dplyr))
```

# The Core tidyverse Transformation Functions

First:

```
library(tidyverse)
```

```
(or just library(dplyr))
```

- ▶ filter()
- ▶ arrange()
- ▶ group\_by() (and ungroup())
- ▶ select() (and rename())
- ▶ mutate()
- ▶ transmute()
- ▶ summarise()

## Common Structure

```
verb(df, <conditions or calculations>)
```

## Common Structure

```
verb(df, <conditions or calculations>)
```

Value: a dataframe

## Common Structure

This structure:

dataframe in  $\rightsquigarrow$  dataframe out

enables the pipe: |>

The pipe inserts the previous result as the first argument of the subsequent function.

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```
x |> f(y)
```

is the same as

```
f(x, y)
```

The pipe inserts the previous result as the first argument of the subsequent function.

```
x |> f(y)
```

is the same as

```
f(x, y)
```

(also, ...)

```
x %>% f(y)
```

## The Pipe

- ▶ Suppose we have functions `f()`, `g()`, and `h()`

## The Pipe

- ▶ Suppose we have functions  $f()$ ,  $g()$ , and  $h()$
- ▶ We want to apply  $f()$  to  $x$ , then apply  $g()$  to the output, then  $h()$  to the output of  $g()$ , ...

## The Pipe

- ▶ Suppose we have functions  $f()$ ,  $g()$ , and  $h()$
- ▶ We want to apply  $f()$  to  $x$ , then apply  $g()$  to the output, then  $h()$  to the output of  $g()$ , ...
- ▶  $f(x)$

## The Pipe

- ▶ Suppose we have functions  $f()$ ,  $g()$ , and  $h()$
- ▶ We want to apply  $f()$  to  $x$ , then apply  $g()$  to the output, then  $h()$  to the output of  $g()$ , ...
- ▶  $f(x)$
- ▶  $g(f(x))$

## The Pipe

- ▶ Suppose we have functions  $f()$ ,  $g()$ , and  $h()$
- ▶ We want to apply  $f()$  to  $x$ , then apply  $g()$  to the output, then  $h()$  to the output of  $g()$ , ...
- ▶  $f(x)$
- ▶  $g(f(x))$
- ▶  $h(g(f(x)))$

## The Pipe

- ▶ Suppose we have functions  $f()$ ,  $g()$ , and  $h()$
- ▶ We want to apply  $f()$  to  $x$ , then apply  $g()$  to the output, then  $h()$  to the output of  $g()$ , ...
- ▶  $f(x)$
- ▶  $g(f(x))$
- ▶  $h(g(f(x)))$

## The Pipe

- ▶ Suppose we have functions  $f()$ ,  $g()$ , and  $h()$
- ▶ We want to apply  $f()$  to  $x$ , then apply  $g()$  to the output, then  $h()$  to the output of  $g()$ , ...
- ▶  $f(x)$
- ▶  $g(f(x))$
- ▶  $h(g(f(x)))$

Or, with more assignments,

- ▶  $y \leftarrow f(x)$

# The Pipe

- ▶ Suppose we have functions  $f()$ ,  $g()$ , and  $h()$
- ▶ We want to apply  $f()$  to  $x$ , then apply  $g()$  to the output, then  $h()$  to the output of  $g()$ , ...
- ▶  $f(x)$
- ▶  $g(f(x))$
- ▶  $h(g(f(x)))$

Or, with more assignments,

- ▶  $y \leftarrow f(x)$
- ▶  $z \leftarrow g(y)$

# The Pipe

- ▶ Suppose we have functions  $f()$ ,  $g()$ , and  $h()$
- ▶ We want to apply  $f()$  to  $x$ , then apply  $g()$  to the output, then  $h()$  to the output of  $g()$ , ...
- ▶  $f(x)$
- ▶  $g(f(x))$
- ▶  $h(g(f(x)))$

Or, with more assignments,

- ▶  $y \leftarrow f(x)$
- ▶  $z \leftarrow g(y)$
- ▶  $h(z)$

## The Pipe

The pipe (`|>`) allows us to write

```
x |> f() |> g() |> h()
```

## The Pipe

The pipe (`|>`) allows us to write

```
x |> f() |> g() |> h()
```

Likely better,

```
x |>  
  f() |>  
  g() |>  
  h()
```

## The Pipe

The pipe (`|>`) allows us to write

```
x |> f() |> g() |> h()
```

Likely better,

```
x |>  
  f() |>  
  g() |>  
  h()
```

To be able to reorder *depends* on functions all

- ▶ taking same first input
- ▶ producing output of same type as input

## The Pipe

The `|>` is like `o` for function composition, but still reads in order.

## The Pipe

The  $|>$  is like  $\circ$  for function composition, but still reads in order.

(Unlike  $h(g(f(x)))$  or  $(h \circ g \circ f)(x)$ )

## The Pipe

The  $|>$  is like  $\circ$  for function composition, but still reads in order.

(Unlike  $h(g(f(x)))$  or  $(h \circ g \circ f)(x)$ )

Read “then”.

# The Pipe

Suppose each function takes more than 1 argument:

## The Pipe

Suppose each function takes more than 1 argument:

```
h(g(f(x, arg1 = value_here), arg2 = another_val),  
  arg3 = 5, arg4 = TRUE)
```

## The Pipe

Suppose each function takes more than 1 argument:

```
h(g(f(x, arg1 = value_here), arg2 = another_val),  
  arg3 = 5, arg4 = TRUE)
```

Messy. Which function is arg2? arg3?

# The Pipe

Suppose each function takes more than 1 argument:

```
h(g(f(x, arg1 = value_here), arg2 = another_val),  
  arg3 = 5, arg4 = TRUE)
```

Messy. Which function is arg2? arg3?

```
x |>  
f(arg1 = value_here) |>  
g(arg2 = another_val) |>  
h(arg3 = 5, arg4 = TRUE)
```

# The Pipe

Suppose each function takes more than 1 argument:

```
h(g(f(x, arg1 = value_here), arg2 = another_val),  
  arg3 = 5, arg4 = TRUE)
```

Messy. Which function is arg2? arg3?

```
x |>  
f(arg1 = value_here) |>  
g(arg2 = another_val) |>  
h(arg3 = 5, arg4 = TRUE)
```

Better.

## The Pipe

Fun note: The `%>%` pipe is defined in package  
`magrittr`

## The Pipe

Fun note: The `%>%` pipe is defined in package  
`magrittr`

The motif is played **all** the way out:  
<http://j.mp/2Eu679T>

## The Pipe

Fun note: The `%>%` pipe is defined in package  
`magrittr`

The motif is played **all** the way out:

<http://j.mp/2Eu679T>

(For similar missing data example, see Amelia.)

## Core Transformation Funcs: Social Pressure Experiment

```
# URL is "https://raw.githubusercontent.com/kosukeimai/
#           qss/master/CAUSALITY/social.csv"
social <- read_csv("http://j.mp/2Et71U0")
dim(social)
```

```
## [1] 305866      6
```

# Core Transformation Funcs: Social Pressure Experiment

```
# URL is "https://raw.githubusercontent.com/kosukeimai/
#           qss/master/CAUSALITY/social.csv"
social <- read_csv("http://j.mp/2Et71U0")
dim(social)
```

```
## [1] 305866      6
```

```
head(social, 4)
```

```
## # A tibble: 4 x 6
##   sex     yearofbirth primary2004 messages    primary2006 hhszie
##   <chr>     <dbl>       <dbl> <chr>        <dbl>    <dbl>
## 1 male      1941         0 Civic Duty      0      2
## 2 female    1947         0 Civic Duty      0      2
## 3 male      1951         0 Hawthorne      1      3
## 4 female    1950         0 Hawthorne      1      3
```

## filter()

Keep only voters in households that might have interference:

## filter()

Keep only voters in households that might have interference:

```
table(social$hhszie)
```

```
##  
##      1      2      3      4      5      6      7  
## 42524 190294 51057 18596 2955 390 42
```

## filter()

Keep only voters in households that might have interference:

```
table(social$hhszie)
```

```
##  
##      1      2      3      4      5      6      7  
## 42524 190294 51057 18596 2955 390 42
```

```
df_interf <- filter(social, hhszie > 1)  
dim(df_interf)
```

```
## [1] 263342       6
```

## filter()

Keep only non-voters who might be subject to interference:

## filter()

Keep only non-voters who might be subject to interference:

```
filter(social, (hhszie > 1) & (primary 2004 == 0))
```

## filter()

Keep only non-voters who might be subject to interference:

```
filter(social, (hhszie > 1) & (primary 2004 == 0))
```

```
## Error in parse(text = input): <text>:1:40: unexpected character: ^  
## 1: filter(social, (hhszie > 1) & (primary 2004  
##
```

## filter()

Keep only non-voters who might be subject to interference:

```
filter(social, (hhszie > 1) & (primary 2004 == 0))
```

```
## Error in parse(text = input): <text>:1:40: unexpected character: ^  
## 1: filter(social, (hhszie > 1) & (primary 2004  
##
```

```
filter(social, (hhszie > 1) & (primary2004 == 0))
```

```
## # A tibble: 161,275 x 6  
##   sex     yearofbirth primary2004 messages primary2006 hhszie  
##   <chr>        <dbl>      <dbl> <chr>       <dbl>    <dbl>  
## 1 male         1941        0 Civic Duty        0  
## 2 female       1947        0 Civic Duty        0  
## 3 male         1951        0 Hawthorne       1  
## 4 female       1950        0 Hawthorne       1  
## 5 female       1982        0 Hawthorne       1  
## 6 male         1981        0 Control          0  
## 7 female       1959        0 Control          0
```

## arrange()

Sort by birth year, then household size

## arrange()

Sort by birth year, then household size

```
social |> arrange(yearofbirth, hhszie)
```

```
## # A tibble: 305,866 x 6
##   sex    yearofbirth primary2004 messages primary2006 hhszie
##   <chr>     <dbl>      <dbl> <chr>      <dbl>    <dbl>
## 1 female     1900        0 Control       0        1
## 2 female     1900        0 Control       0        2
## 3 male       1900        1 Control       0        2
## 4 male       1900        1 Control       1        2
## 5 female     1900        0 Hawthorne    0        2
## 6 female     1900        1 Control       1        3
## 7 female     1902        1 Control       0        1
## 8 female     1902        1 Control       0        3
## 9 male       1903        1 Control       0        1
## 10 female    1904        0 Control      0        1
## # i 305,856 more rows
```

## `mutate()`

Create new variable (`under_30`), TRUE/FALSE

## mutate()

Create new variable (under\_30), TRUE/FALSE

```
social |>
  mutate(under_30 = yearofbirth > 1976) |>
  select(sex, yearofbirth, under_30)
```

```
## # A tibble: 305,866 x 3
##   sex   yearofbirth under_30
##   <chr>     <dbl>   <lgl>
## 1 male      1941 FALSE
## 2 female    1947 FALSE
## 3 male      1951 FALSE
## 4 female    1950 FALSE
## 5 female    1982 TRUE
## 6 male      1981 TRUE
## 7 female    1959 FALSE
## 8 male      1956 FALSE
## 9 female    1968 FALSE
## 10 male     1967 FALSE
## # i 305,856 more rows
```

## mutate()

Create new variable (under\_30), TRUE/FALSE

```
social |>
  mutate(under_30 = yearofbirth > 1976) |>
  select(sex, yearofbirth, under_30)
```

```
## # A tibble: 305,866 x 3
##   sex   yearofbirth under_30
##   <chr>     <dbl>   <lgl>
## 1 male      1941 FALSE
## 2 female    1947 FALSE
## 3 male      1951 FALSE
## 4 female    1950 FALSE
## 5 female    1982 TRUE
## 6 male      1981 TRUE
## 7 female    1959 FALSE
## 8 male      1956 FALSE
## 9 female    1968 FALSE
## 10 male     1967 FALSE
## # i 305,856 more rows
```

(There are also `recode()`, `if_else()`, `case_when()`)

## mutate()

Add new variable to dataset:

```
social <- social |> mutate(under_30 = yearofbirth > 1976,  
                           age = 2006 - yearofbirth)
```

```
## # A tibble: 305,866 x 4  
##   sex    yearofbirth under_30    age  
##   <chr>      <dbl> <lgl>     <dbl>  
## 1 male        1941 FALSE      65  
## 2 female      1947 FALSE      59  
## 3 male        1951 FALSE      55  
## 4 female      1950 FALSE      56  
## 5 female      1982 TRUE       24  
## 6 male        1981 TRUE       25  
## 7 female      1959 FALSE      47  
## 8 male        1956 FALSE      50  
## 9 female      1968 FALSE      38  
## 10 male       1967 FALSE      39  
## # i 305,856 more rows
```

`mutate_all()`, `mutate_at()`, `mutate_if()`

```
soc_numeric <- social |> select(-sex, -messages)
```

```
mutate_all(), mutate_at(), mutate_if()
```

```
soc_numeric <- social |> select(-sex, -messages)
```

```
# Halve every column's values:
```

```
divide_by_two <- function(x){x / 2}
```

```
mutate_all(soc_numeric, divide_by_two)
```

```
## # A tibble: 305,866 x 6
```

```
##   yearofbirth primary2004 primary2006 hhszie under_30
```

```
##   <dbl>      <dbl>      <dbl>    <dbl>      <dbl>
```

```
## 1     970.        0        0       1        0
```

```
## 2     974.        0        0       1        0
```

```
## 3     976.        0        0.5     1.5     0
```

```
## 4     975.        0        0.5     1.5     0
```

```
## 5     991.        0        0.5     1.5     0.5
```

```
## 6     990.        0        0       1.5     0.5
```

```
## 7     980.        0        0.5     1.5     0
```

```
## 8     978.        0        0.5     1.5     0
```

```
## 9     984.        0        0       1        0
```

## `mutate_all()`, `mutate_at()`, `mutate_if()`

*# Double values of columns:*

```
mult_by_two <- function(x){x * 2}
mutate_at(soc_numeric, c(2, 3), mult_by_two)
```

```
## # A tibble: 305,866 x 6
##   yearofbirth primary2004 primary2006 hhszie under_30
##   <dbl>        <dbl>        <dbl>    <dbl>    <lgl>
## 1 1941          0           0         2 FALSE
## 2 1947          0           0         2 FALSE
## 3 1951          0           2         3 FALSE
## 4 1950          0           2         3 FALSE
## 5 1982          0           2         3 TRUE 
## 6 1981          0           0         3 TRUE 
## 7 1959          0           2         3 FALSE
## 8 1956          0           2         3 FALSE
## 9 1968          0           0         2 FALSE
## 10 1967         0           0         2 FALSE
## # i 305,856 more rows
```

`mutate_all()`, `mutate_at()`, `mutate_if()`

What does this do?

```
mutate_at(soc_numeric, vars(matches("primary")),
          mult_by_two)
```

## `mutate_all()`, `mutate_at()`, `mutate_if()`

What does this do?

```
mutate_at(soc_numeric, vars(matches("primary")),
          mult_by_two)
```

```
## # A tibble: 305,866 x 6
##   yearofbirth primary2004 primary2006 hhszie under_30
##       <dbl>      <dbl>      <dbl>     <dbl>     <dbl>    <lgl>
## 1      1941        0        0        0        2 FALSE
## 2      1947        0        0        0        2 FALSE
## 3      1951        0        2        2        3 FALSE
## 4      1950        0        2        3        3 FALSE
## 5      1982        0        2        3        3 TRUE 
## 6      1981        0        0        3        3 TRUE 
## 7      1959        0        2        3        3 FALSE
## 8      1956        0        2        3        3 FALSE
## 9      1968        0        0        2        2 FALSE
## 10     1967        0        0        0        2 FALSE
```

```
## # i 305,866 more rows
```

`mutate_all()`, `mutate_at()`, `mutate_if()`

What does this do?

```
mutate_if(social, is.numeric, mean)
```

## `mutate_all()`, `mutate_at()`, `mutate_if()`

What does this do?

```
mutate_if(social, is.numeric, mean)
```

```
## # A tibble: 305,866 x 8
##   sex     yearofbirth primary2004 messages primary2000
##   <chr>        <dbl>      <dbl> <chr>       <dbl>
## 1 male         1956.     0.401 Civic Duty  0.312
## 2 female       1956.     0.401 Civic Duty  0.312
## 3 male         1956.     0.401 Hawthorne 0.312
## 4 female       1956.     0.401 Hawthorne 0.312
## 5 female       1956.     0.401 Hawthorne 0.312
## 6 male         1956.     0.401 Control   0.312
## 7 female       1956.     0.401 Control   0.312
## 8 male         1956.     0.401 Control   0.312
## 9 female       1956.     0.401 Control   0.312
## 10 male        1956.     0.401 Control  0.312
## # i 305,856 more rows
```

`mutate_all()`, `mutate_at()`, `mutate_if()`

Warning: `mutate_all()`, `_at()`, `_if()` **overwrite** columns  
that are processed.

`mutate_all()`, `mutate_at()`, `mutate_if()`

Warning: `mutate_all()`, `_at()`, `_if()` **overwrite** columns  
that are processed.

Do **not** append new columns to the end.

`mutate_all()`, `mutate_at()`, `mutate_if()`

Warning: `mutate_all()`, `_at()`, `_if()` **overwrite** columns that are processed.

Do **not** append new columns to the end.

Useful for recoding, if want values of a function:

```
is_CD <- function(x){ x == "Civic Duty"}  
mutate_at(social, vars(matches("messages"))), is_CD)
```

## `mutate_all()`, `mutate_at()`, `mutate_if()`

Warning: `mutate_all()`, `_at()`, `_if()` **overwrite** columns that are processed.

Do **not** append new columns to the end.

Useful for recoding, if want values of a function:

```
is_CD <- function(x){ x == "Civic Duty"}  
mutate_at(social, vars(matches("messages"))), is_CD
```

```
## # A tibble: 305,866 x 8  
##   sex     yearofbirth primary2004 messages primary2006 hhsiz...  
##   <chr>      <dbl>       <dbl> <lgl>       <dbl>    <dbl>  
## 1 male        1941         0 TRUE          0         2  
## 2 female      1947         0 TRUE          0         2  
## 3 male        1951         0 FALSE         1         3  
## 4 female      1950         0 FALSE         1         3  
## 5 female      1982         0 FALSE         1         3  
## 6 male        1981         0 FALSE         0         3  
## 7 female      1959         0 FALSE         1         3  
## 8 male        1956         0 FALSE        1231 / 2743
```

`mutate()` with `if_else()`, `case_when()`

Create new variable (`age_cat`: "senior", "non-senior"  
using `age`):

## mutate() with if\_else(), case\_when()

Create new variable (age\_cat: "senior", "non-senior" using age):

```
social |>  
  mutate(age_cat = if_else(age >= 55, "older", "not-older")) |>  
  select(sex, yearofbirth, messages, under_30, age, age_cat)
```

```
## # A tibble: 305,866 x 6  
##   sex     yearofbirth messages  under_30    age age_cat  
##   <chr>      <dbl> <chr>     <lgl>     <dbl> <chr>  
## 1 male        1941 Civic Duty FALSE      65 older  
## 2 female      1947 Civic Duty FALSE      59 older  
## 3 male        1951 Hawthorne FALSE      55 older  
## 4 female      1950 Hawthorne FALSE      56 older  
## 5 female      1982 Hawthorne TRUE       24 not-older  
## 6 male        1981 Control    TRUE       25 not-older  
## 7 female      1959 Control    FALSE      47 not-older  
## 8 male        1956 Control    FALSE      50 not-older  
## 9 female      1968 Control    FALSE      38 not-older  
## 10 male       1967 Control    FALSE      39 not-older  
## # i 305,856 more rows
```

`mutate()` with `ifelse()`, `case_when()`

Create new variable (`age_cat_3` using `age`):

## mutate() with ifelse(), case\_when()

Create new variable (age\_cat\_3 using age):

```
social |>  
  mutate(age_cat_3 = case_when(  
    age < 30 ~ "young",  
    age > 55 ~ "older",  
    TRUE ~ "middle")) |>  
  select(sex, yearofbirth, messages, age, age_cat_3)
```

```
## # A tibble: 305,866 x 5  
##   sex     yearofbirth messages     age age_cat_3  
##   <chr>      <dbl> <chr>     <dbl> <chr>  
## 1 male        1941 Civic Duty     65 older  
## 2 female      1947 Civic Duty     59 older  
## 3 male        1951 Hawthorne     55 middle  
## 4 female      1950 Hawthorne     56 older  
## 5 female      1982 Hawthorne     24 young  
## 6 male         1981 Control      25 young  
## 7 female      1959 Control      47 middle  
## 8 male         1956 Control      50 middle  
## 9 female      1968 Control      38 middle  
## 10 male        1967 Control      39 middle  
## # i 305,856 more rows
```

`transmute()` for new variables, summaries

```
social |> transmute(age = 2006 - yearofbirth)
```

## `transmute()` for new variables, summaries

```
social |> transmute(age = 2006 - yearofbirth)
```

```
## # A tibble: 305,866 x 1
##       age
##   <dbl>
## 1     65
## 2     59
## 3     55
## 4     56
## 5     24
## 6     25
## 7     47
## 8     50
## 9     38
## 10    39
## # i 305,856 more rows
```

`transmute()` for new vars, summaries as new vars

```
social_msg_grps <- group_by(social, messages)
```

`transmute()` for new vars, summaries as new vars

```
social_msg_grps <- group_by(social, messages)
```

```
social_msg_grps |>  
  transmute(avg_age = mean(2006 - yearofbirth))
```

```
## # A tibble: 305,866 x 2  
## # Groups:   messages [4]  
##       messages     avg_age  
##       <chr>        <dbl>  
## 1 Civic Duty      49.7  
## 2 Civic Duty      49.7  
## 3 Hawthorne       49.7  
## 4 Hawthorne       49.7  
## 5 Hawthorne       49.7  
## 6 Control          49.8  
## 7 Control          49.8  
## 8 Control          49.8  
## 9 Control          49.8
```

What if I wanted just mean age per message?

What if I wanted just mean age per message?

```
social_msg_grps |>  
  summarise(avg_age = mean(2006 - yearofbirth))
```

```
## # A tibble: 4 x 2  
##   messages     avg_age  
##   <chr>        <dbl>  
## 1 Civic Duty    49.7  
## 2 Control       49.8  
## 3 Hawthorne     49.7  
## 4 Neighbors      49.9
```

What if I wanted just mean age per message?

```
social_msg_grps |>  
  summarise(avg_age = mean(2006 - yearofbirth))
```

```
## # A tibble: 4 x 2  
##   messages     avg_age  
##   <chr>        <dbl>  
## 1 Civic Duty    49.7  
## 2 Control       49.8  
## 3 Hawthorne     49.7  
## 4 Neighbors      49.9
```

What information does this provide about the experiment?

`select()`

```
select(social, yearofbirth, messages, primary2006) # or  
social |> select(yearofbirth, messages, primary2006)
```

## select()

```
select(social, yearofbirth, messages, primary2006) # or  
social |> select(yearofbirth, messages, primary2006)
```

```
## # A tibble: 305,866 x 3  
##   yearofbirth messages  primary2006  
##       <dbl>    <chr>        <dbl>  
## 1      1941 Civic Duty      0  
## 2      1947 Civic Duty      0  
## 3      1951 Hawthorne      1  
## 4      1950 Hawthorne      1  
## 5      1982 Hawthorne      1  
## 6      1981 Control        0  
## 7      1959 Control        1  
## 8      1956 Control        1  
## 9      1968 Control        0  
## 10     1967 Control        0  
## # i 305,856 more rows
```

## Other Common Transformation Functions

## Other Common Transformation Functions: `slice()`

```
slice(social, 1000:1004)
```

```
## # A tibble: 5 x 8
##   sex     yearofbirth primary2004 messages primary2006 l
##   <chr>        <dbl>      <dbl> <chr>      <dbl>
## 1 male         1955       1 Neighbors    1
## 2 female       1952       0 Control     1
## 3 male         1947       1 Control     1
## 4 female       1985       0 Hawthorne   0
## 5 male         1956       0 Hawthorne   0
```

## Other Common Transformation Functions: `slice()`

```
slice(social, n())
```

```
## # A tibble: 1 x 8
##   sex      yearofbirth primary2004 messages primary2006 h...
##   <chr>        <dbl>       <dbl> <chr>       <dbl> ...
## 1 female      1949        1 Control     1
```

## Other Common Transformation Functions: sample\_n(), sample\_frac()

```
sample_n(social, 4)
```

```
## # A tibble: 4 x 8
##   sex     yearofbirth primary2004 messages primary2006 h...
##   <chr>      <dbl>       <dbl> <chr>       <dbl> ...
## 1 male        1965        0 Control       0
## 2 male        1930        1 Control       0
## 3 female      1953        0 Control       0
## 4 female      1959        0 Control       0
```

## Other Common Transformation Functions: sample\_n(), sample\_frac()

```
sample_frac(social, 0.00001)
```

```
## # A tibble: 3 x 8
##   sex   yearofbirth primary2004 messages primary2006 hh
##   <chr>     <dbl>      <dbl> <chr>      <dbl> 
## 1 male       1956        0 Neighbors      0
## 2 male       1953        0 Control        0
## 3 male       1963        0 Control        1
```

## Other Common Transformation Functions: `distinct()`

```
social_distinct <- distinct(social)
dim(social_distinct)
```

```
## [1] 9235     8
```

## Other Common Transformation Functions: `distinct()`

```
social_distinct <- distinct(social)
dim(social_distinct)
```

```
## [1] 9235     8
```

$$(100 \text{ yrs}) \cdot (4 \text{ msgs}) \cdot (4 \text{ votes}) \cdot (2 \text{ sex}) \cdot (3 \text{ HHsize}) = 9600$$

## Helper Functions

## Helpers for `select()`-ing Variables

- ▶ `contains()`
- ▶ `starts_with()`, `ends_with()`
- ▶ `matches()`
- ▶ `num_range()`
- ▶ `one_of()`
- ▶ `everything()`

## Helpers for `select()`-ing Variables

```
social |> select(contains("s")) |> slice(1:2)
```

```
## # A tibble: 2 x 3
##   sex     messages    hhszie
##   <chr>   <chr>       <dbl>
## 1 male    Civic Duty      2
## 2 female  Civic Duty      2
```

```
# literal string "s"
```

## Helpers for `select()`-ing Variables

```
social |> select(contains("s")) |> slice(1:2)
```

```
## # A tibble: 2 x 3
##   sex     messages    hhszie
##   <chr>   <chr>       <dbl>
## 1 male    Civic Duty      2
## 2 female  Civic Duty      2
```

```
# literal string "s"
```

```
social |> select(starts_with("primary")) |> slice(1:2)
```

```
## # A tibble: 2 x 2
##   primary2004 primary2006
##   <dbl>        <dbl>
## 1 0            0
## 2 0            0
```

## Helpers for `select()`-ing Variables

```
social |> select(ends_with("size")) |> slice(1:2)
```

```
## # A tibble: 2 x 1
##   hhszie
##   <dbl>
## 1     2
## 2     2
```

## Helpers for `select()`-ing Variables

```
social |> select(ends_with("size")) |> slice(1:2)
```

```
## # A tibble: 2 x 1
##   hhszie
##   <dbl>
## 1     2
## 2     2
```

```
social |> select(matches(".00.")) |> slice(1:2) # regex
```

```
## # A tibble: 2 x 2
##   primary2004 primary2006
##   <dbl>        <dbl>
## 1     0          0
## 2     0          0
```

## Helpers for `select()`-ing Variables

```
social |> select(num_range("primary", 2000:2008)) |> slice(1)
```

```
## # A tibble: 2 x 2
##   primary2004 primary2006
##       <dbl>        <dbl>
## 1         0          0
## 2         0          0
```

## Helpers for `select()`-ing Variables

```
social |> select(num_range("primary", 2000:2008)) |> slice(1)
```

```
## # A tibble: 2 x 2
##   primary2004 primary2006
##       <dbl>        <dbl>
## 1         0          0
## 2         0          0
```

But

```
social |> select(num_range("primary", 2000:2005)) |> slice(1)
```

```
## # A tibble: 2 x 1
##   primary2004
##       <dbl>
## 1         0
## 2         0
```

## Helpers for `select()`-ing Variables

```
social |> select(one_of(c("sex", "height"))) |> slice(1:5)

## # A tibble: 5 x 1
##   sex
##   <chr>
## 1 male
## 2 female
## 3 male
## 4 female
## 5 female
```

## Helpers for `select()`-ing Variables

```
social |> select(primary2006, messages, everything()) |>  
  slice(1:9)
```

```
## # A tibble: 9 x 8  
##   primary2006 messages   sex   yearofbirth primary2004  
##   <dbl>     <chr>    <chr>     <dbl>      <dbl>  
## 1 0        Civic Duty male    1941       0  
## 2 0        Civic Duty female  1947       0  
## 3 1        Hawthorne male    1951       0  
## 4 1        Hawthorne female   1950       0  
## 5 1        Hawthorne female   1982       0  
## 6 0        Control   male    1981       0  
## 7 1        Control   female   1959       0  
## 8 1        Control   male    1956       0  
## 9 0        Control   female   1968       0
```

## Helpers for `select()`-ing Variables

```
social |> select(primary2006, messages, everything()) |>  
  slice(1:9)
```

```
## # A tibble: 9 x 8  
##   primary2006 messages   sex   yearofbirth primary2004  
##   <dbl>     <chr>     <chr>    <dbl>      <dbl>  
## 1 0        Civic Duty male    1941       0  
## 2 0        Civic Duty female  1947       0  
## 3 1        Hawthorne male    1951       0  
## 4 1        Hawthorne female   1950       0  
## 5 1        Hawthorne female   1982       0  
## 6 0        Control   male    1981       0  
## 7 1        Control   female   1959       0  
## 8 1        Control   male    1956       0  
## 9 0        Control   female   1968       0
```

(Use `select()` as the `arrange()` of columns.)

## Helpers for `mutate()`

- ▶ Offsets
- ▶ Cumulative aggregates
- ▶ Ranking functions

## Viewing the Data

- ▶ df
- ▶ View(df)
- ▶ as.data.frame(tbl)
- ▶ tbl |> as.data.frame()

## Recently, at The Lab... preprocessing

```
df_outcomes <- df_outcomes |>  
  rename(ic_case_id = "IC# (Household)",  
         pdc_number = "TANF PDC Case #",  
         pdc_status = "PDC Current Status",  
         renewal_date = "Renewal Date"  
    )
```

## Recently, at The Lab... preprocessing

```
df_outcomes <- df_outcomes |>  
  rename(ic_case_id = "IC# (Household)",  
         pdc_number = "TANF PDC Case #",  
         pdc_status = "PDC Current Status",  
         renewal_date = "Renewal Date"  
     )
```

(though, see `janitor::clean_names()`)

## Recently, at The Lab...

```
df_arrest <- df_arrest |>  
  rename(age = Age,  
         race = `Defendant Race`)  
  
df_stop <- df_stop |>  
  rename(age = `Subject Age`,  
         race = `Subject_Race`)
```

## Recently, at The Lab...

```
df_stop$sex <- recode(df_stop$sex,  
                      Female = "F", Male = "M")  
  
df_stop$sex <- na_if(df_stop$sex, "Unknown")
```

## Recently, at The Lab... deduplication

```
final_baseline_data <- final_baseline_data |>  
  filter(!((ic_case_id == 1234) & (pdc_number == 2))) |>  
  filter(!((ic_case_id == 5678) &  
           (address == "1600 Pennsylvania Ave NW"))) |>  
  filter(!((ic_case_id == 6961) & (pdc_number == 9))) |>  
  filter(!((ic_case_id == 2087) & (pdc_number == 7)))
```

## Recently, at The Lab...

```
df_only_dup_months <- df_only_duplicated |>
  group_by(ic_case_id) |>
  summarise(month_count = n_distinct(recert_month)) |>
  filter(month_count > 1) |>
  select(ic_case_id)
```

# Comparing Base R vs. the Tidyverse

Which do you prefer?

```
df [1, 3]
```

vs.

```
df |>  
  slice(1) |>  
  select(3)
```

# Comparing Base R vs. the Tidyverse

Which do you prefer?

```
select(df, x1, x2)
```

vs.

```
df |> select(x1, x2)
```

vs.

```
df[, c("x1", "x2")]
```

# Core Transformation Functions Quiz

Suppose we have dataframe `df` with 100 rows, continuous variable `x` and categorical `y`.

Hand-write code<sup>1</sup> to

1. sort `df` by the values of `x`? (largest first)
2. create a new variable `x_sq` – the square of each row's `x` value – and attach it as a column of `df`?
3. create `df2`, which has only the rows of `df` where `x > 5`?
4. calculate the median value of `x` within categories of `y`?

---

<sup>1</sup>`filter()`, `arrange()`, `group_by()`, `ungroup()`, `select()`, `rename()`, `mutate()`, `transmute()`, `summarise()`

## References

Wickham, Hadley, Mine Çetinkaya-Rundel, and Garrett Grolemund. 2023. *R for Data Science: Import, Tidy, Transform, Visualize, and Model Data*. 2nd ed. O'Reilly Media. <http://r4ds.hadley.nz/>.