### Functions and iteration

#### Week 12

AEM 2850 / 5850 : R for Business Analytics Cornell Dyson Spring 2025

Acknowledgements: Claus Wilke

### **Announcements**

### **Group project due this Friday, April 18!**

Office hours for the rest of this week:

- Tuesday: Prof. Gerarden open office hours from 11:30-12:30 in Warren 464
- Tuesday: Prof. Gerarden by appointment at aem2850.youcanbook.me
- Thursday: Prof. Gerarden by appointment (already fully booked)
- Friday: extra TA office hours from 2:00-4:00 in Warren 175

We will have a regular homework this week (due Monday 4/21)

We will have regular TA office hours on Monday

Questions before we get started?

### Plan for this week

### **Tuesday**

- Intro to functions and iteration
- example-12-1

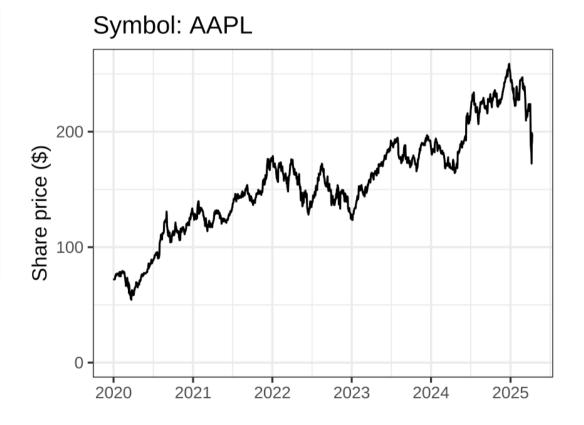
### **Thursday**

- Conditional execution
- Functions with multiple arguments
- example-12-2

### Intro to functions and iteration

### We often run similar code multiple times

What needs to change if we want to look at AMZN share prices instead?



### We often run similar code multiple times



### We often run similar code multiple times



### How can we avoid duplication and mistakes?

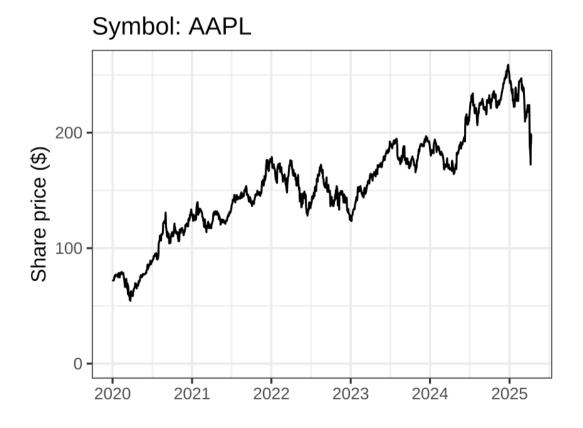
- 1. Avoid hard-coding specific values
- 2. Define a function
- 3. Automate calling the function
- 4. Write a more general function
- 5. Use these concepts in a tidy pipeline

We will focus on steps 1-3 due to time constraints

What is "hard-coded" here?

How can we avoid this hard-coding?

str\_glue() allows us to put the contents
of ticker in the plot's title



Now **ticker** is the only thing that changes



Now **ticker** is the only thing that changes



### Step 2: Define a function

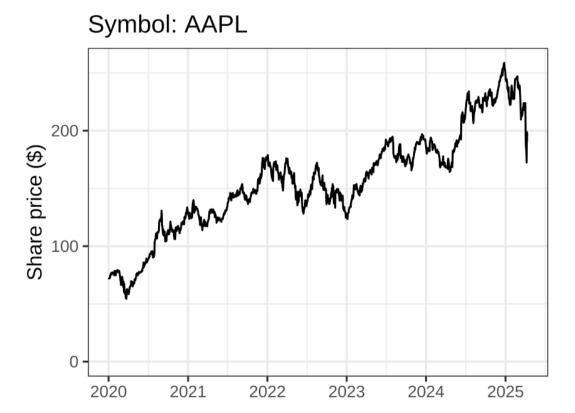
### Three key steps:

- 1. Pick a **name**
- 2. List **arguments** inside function()
- 3. Put code in the **body** of the function, delimited by { . . . }

Easiest to write the body on a test case, *then* convert it into a function

### Step 2: Define a function

```
make_plot <- function(ticker) {</pre>
  sp500_prices |>
    filter(symbol == ticker) |>
    ggplot(aes(x = date, y = adjusted)) +
    geom_line() +
    labs(x = NULL,
         y = "Share price ($)",
         title = str_glue("Symbol: {ticker}"))
    scale_x_date(date_breaks = "1 year",
                 date labels = "%Y") +
    scale_y_continuous(limits = c(0, NA)) +
    theme_bw()
make_plot("AAPL")
```



### Step 2: Define a function

make\_plot("AMZN")



make\_plot("TSLA")



### Rules of thumb about functions

- You can (almost) never write too many functions
- When you find yourself writing the same code 3+ times, put it into a function
- A function should be no longer than 20-40 lines
- If a function is getting too long, break it into smaller functions

Individual function calls are hard to scale

```
make_plot("AAPL")
make_plot("AMZN")
make_plot("TSLA")
```

What if we wanted to make this plot for every company in the S&P 500?

How could you automate these function calls?

- 1. Imperative programming (for loops)
- 2. Functional programming (map functions)

The purrr packages provides map functions that take a vector as input, apply a **function** to each element of the vector, and return the results in a new vector:

```
map(some_vector, some_function)
```

map functions are basically identical to base R's apply functions

### How can we use map to make plots for AAPL, AMZN, and TSLA?

```
symbols <- c("AAPL", "AMZN", "TSLA")
plots <- map(symbols, make_plot)</pre>
```

Here map takes each element of the vector symbols and uses it as input for our function make\_plot()

map returns a **list**. In this example, it's a list of plots that we assigned to plots:

class(plots)

## [1] "list"

plots[[1]]



plots[[2]]



The syntax plots [[x]] allows us to drill down into the list plots and extract whatever object is in the xth position (here: a ggplot)

### This scales really easily!

all\_symbols <- sp500\_prices |> distinct(symbol) |> pull() # get all the symbols in the S&P 500 all\_plots <- map(all\_symbols, make\_plot) # make a plot for each of the symbols

length(all\_symbols)

## [1] 505

length(all\_plots)

## [1] 505

all\_plots[[35]]

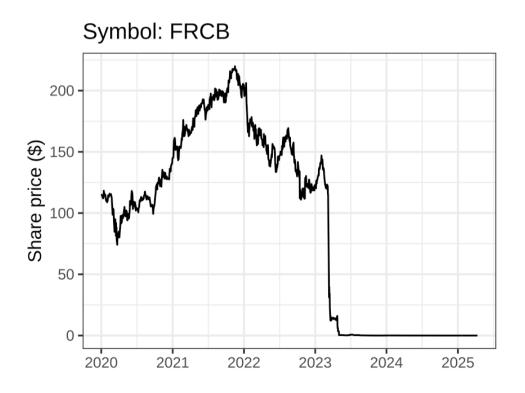


all\_plots[[500]]



We can also extract results using logical expressions:

```
all_plots[all_symbols=="FRCB"]
```



### The map functions

The purrr package provides a family of map functions that return different types of output:

- map() makes a list
- map\_lgl() makes a logical vector
- map\_int() makes an integer vector
- map\_dbl() makes a double vector
- map\_chr() makes a character vector

## What about for loops?

#### For loops work too!

```
symbols <- c("AAPL", "AMZN", "TSLA")
plots <- vector("list", length(symbols)) # 1. allocate space for output
for (i in seq_along(symbols)) { # 2. specify the sequence to loop over
   plots[[i]] <- make_plot(symbols[i]) # 3. specify what to do in each iteration
}</pre>
```

#### But functional programming is more concise:

```
symbols <- c("AAPL", "AMZN", "TSLA")
plots <- map(symbols, make_plot)</pre>
```

### Why not use for loops?

- They often require us to think about data logistics (indexing)
- They encourage iterative thinking over conceptual thinking
- Typically require more code, which often means more errors
- Can be harder to parallelize or otherwise optimize

### But there is nothing wrong with using them!

We can practice using for loops during the example if time permits

# example-12-1

### Conditional execution

# Functions with multiple arguments

# example-12-2