

# Functions and iteration

**Week 12**

AEM 2850 / 5850 : R for Business Analytics  
Cornell Dyson  
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Acknowledgements: **Claus Wilke**

# Announcements

## **Group project due this Friday, November 14!**

Office hours for the rest of this week:

- Tuesday: Prof. Gerarden open office hours from 11:30-12:30 in Warren 464
- Thursday: Prof. Gerarden by appointment at [aem2850.youcanbook.me](http://aem2850.youcanbook.me)
- Friday: TA office hours from 1:00-2:00 in Warren B05

We will have a homework this week (due Monday 11/17)

- Will work through part of the homework on Thursday in class
- 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

## Thursday

- Conditional execution
- Functions with multiple arguments
- homework-12

# Intro to functions and iteration

# We often run similar code multiple times

```
sp500_prices |>  
  filter(symbol == "AAPL") |>  
  ggplot(aes(x = date, y = adjusted)) +  
  geom_line() +  
  labs(x = NULL,  
       y = "Share price ($)",  
       title = "Symbol: AAPL") +  
  scale_x_date(date_breaks = "1 year",  
               date_labels = "%Y") +  
  scale_y_continuous(limits = c(0, NA)) +  
  theme_bw()
```

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

Symbol: AAPL



# We often run similar code multiple times

```
sp500_prices |>  
  filter(symbol == "AMZN") |>  
  ggplot(aes(x = date, y = adjusted)) +  
  geom_line() +  
  labs(x = NULL,  
       y = "Share price ($)",  
       title = "Symbol: AMZN") +  
  scale_x_date(date_breaks = "1 year",  
               date_labels = "%Y") +  
  scale_y_continuous(limits = c(0, NA)) +  
  theme_bw()
```



# We often run similar code multiple times

```
sp500_prices |>  
  filter(symbol == "TSLA") |>  
  ggplot(aes(x = date, y = adjusted)) +  
  geom_line() +  
  labs(x = NULL,  
       y = "Share price ($)",  
       title = "Symbol: TSLA") +  
  scale_x_date(date_breaks = "1 year",  
               date_labels = "%Y") +  
  scale_y_continuous(limits = c(0, NA)) +  
  theme_bw()
```



# 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

# Step 1: Avoid hard-coding specific values

```
sp500_prices |>
  filter(symbol == "AAPL") |>
  ggplot(aes(x = date, y = adjusted)) +
  geom_line() +
  labs(x = NULL,
       y = "Share price ($)",
       title = "Symbol: AAPL") +
  scale_x_date(date_breaks = "1 year",
               date_labels = "%Y") +
  scale_y_continuous(limits = c(0, NA)) +
  theme_bw()
```

What is "hard-coded" here?

# Step 1: Avoid hard-coding specific values

```
sp500_prices |>
  filter(symbol == "AAPL") |>
  ggplot(aes(x = date, y = adjusted)) +
  geom_line() +
  labs(x = NULL,
       y = "Share price ($)",
       title = "Symbol: AAPL") +
  scale_x_date(date_breaks = "1 year",
               date_labels = "%Y") +
  scale_y_continuous(limits = c(0, NA)) +
  theme_bw()
```

How can we avoid this hard-coding?

# Step 1: Avoid hard-coding specific values

```
ticker <- "AAPL"

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()
```

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

Symbol: AAPL



# Step 1: Avoid hard-coding specific values

```
ticker <- "AMZN"

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()
```

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



# Step 1: Avoid hard-coding specific values

```
ticker <- "TSLA"

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()
```

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



# Step 2: Define a function

```
make_plot <- function(ticker) {  
  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()  
}
```

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) {  
  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")
```

Symbol: 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

# Step 3: Automate calling the function

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)**

# Step 3: Automate calling the function

The `purrr` package 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)
```

Here `map` takes each element of the vector `symbols` and uses it as input for our function `make_plot()`

# Step 3: Automate calling the function

`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 `x`th position (here: a ggplot)

# Step 3: Automate calling the function

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]]
```



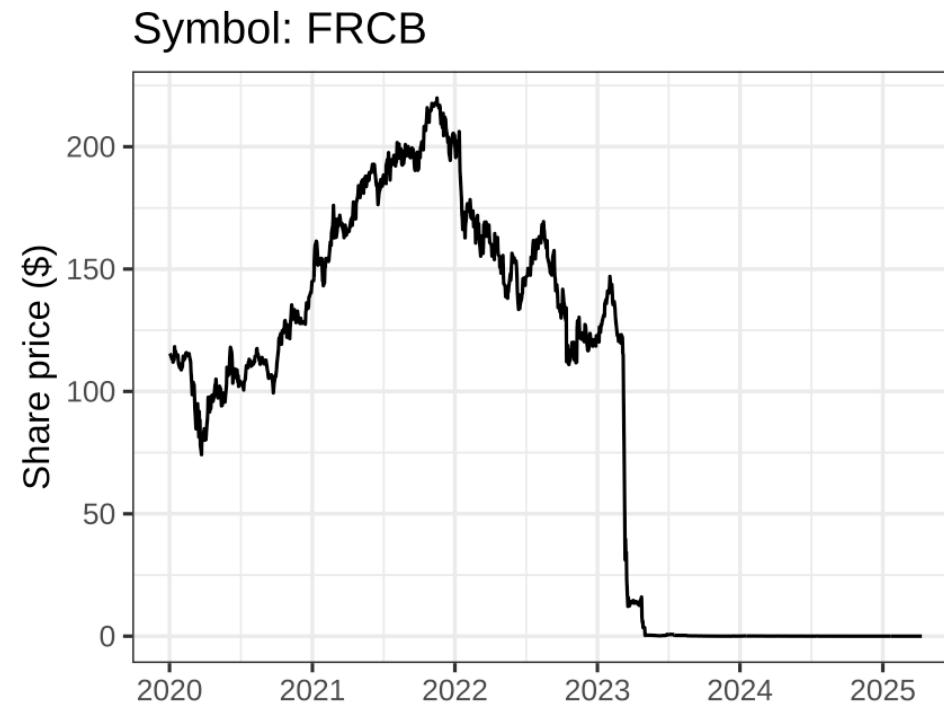
```
all_plots[[500]]
```



# Step 3: Automate calling the function

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
}
```

But functional programming is more concise:

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

# 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**

# Conditional execution

# Functions with multiple arguments

homework-12