

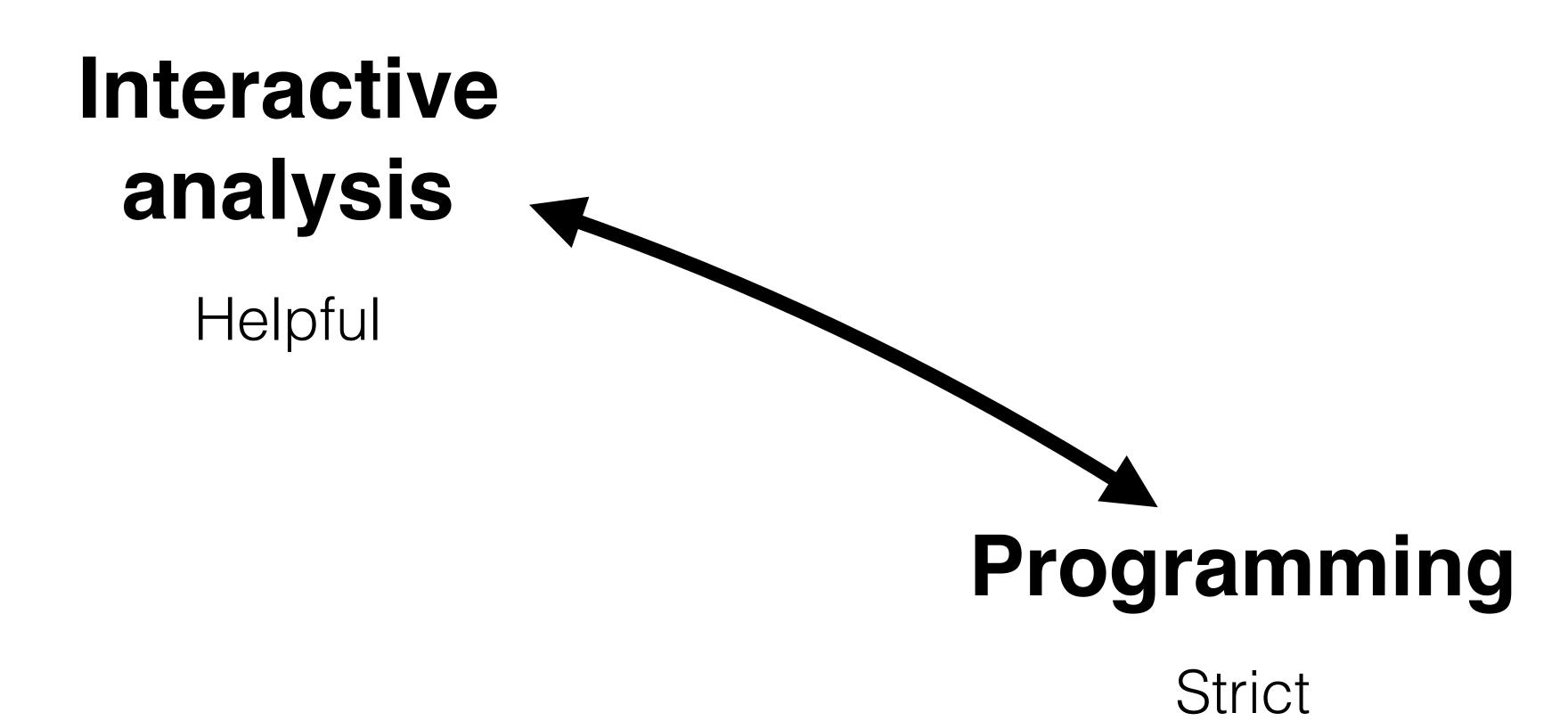


Robust code

What do these calls do?

```
> df[, vars]
> subset(df, x == y)
> data.frame(x = "a")
```







Three main problems

- Type-unstable functions
- Non-standard evaluation
- Hidden arguments



Throwing errors

```
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> stopifnot(is.character(x))
Error: is.character(x) is not TRUE
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}
```

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```
if (condition) {
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}
```

```
> if (!is.character(x)) {
    stop("`x` should be a character vector", call. = FALSE)
  }
Error: `x` should be a character vector
```





Let's practice!





Unstable types



Surprises due to unstable types

- Type-inconsistent: the type of the return object depends on the input
- Surprises occur when you've used a type-inconsistent function inside your own function
- Sometimes lead to hard to decipher error messages

What will df[1,] return?

```
> df <- data.frame(z = 1:3, y = 2:4)
> str(df[1, ])
'data.frame':1 obs. of 2 variables:
   $ z: int 1
   $ y: int 2

> df <- data.frame(z = 1:3)
> str(df[1, ])
   int 1
```



[is a common source of surprises

```
> last_row <- function(df) {
        df[nrow(df), ]
   }

> df <- data.frame(x = 1:3)

# Not a row, just a vector
> str(last_row(df))
   int 3
```

Two common solutions for [

```
> last_row <- function(df) {
    df[nrow(df), , drop = FALSE]
  }
> df <- data.frame(x = 1:3)
> str(last_row(df))
'data.frame':1 obs. of 1 variable:
  $ x: int 3
```

- Usedrop = FALSE: df[x,, drop = FALSE]
- Subset the data frame like a list: df [x]

What to do?

- Write your own functions to be type-stable
- Learn the common type-inconsistent functions in R:
 [, sapply
- Avoid using type-inconsistent functions inside your own functions
- Build a vocabulary of type-consistent functions





Let's practice!





Non-standard evaluation

What is non-standard evaluation?

- Non-standard evaluation functions don't use the usual lookup rules
- Great for data analysis, because they save typing

Other NSE functions

```
> library(ggplot2)
> ggplot(mpg, aes(displ, cty)) + geom_point()
> library(dplyr)
> filter(mtcars, disp > 400)
   mpg cyl disp hp drat wt qsec vs am gear carb

      1 10.4
      8 472 205 2.93 5.250 17.98 0 0 3 4

      2 10.4
      8 460 215 3.00 5.424 17.82 0 0 3 4

      3 14.7
      8 440 230 3.23 5.345 17.42 0 0 3 4

> disp_threshold <- 400</pre>
> filter(mtcars, disp > disp_threshold)
    mpg cyl disp hp drat wt qsec vs am gear carb
            8 472 205 2.93 5.250 17.98
2 10.4
               460 215 3.00 5.424 17.82 0 0
         8 440 230 3.23 5.345 17.42 0 0
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> disp_threshold <- 400</pre>
                                                     disp_threshold value in
> filter(mtcars, disp > disp_threshold) the global environment
    mpg cyl disp hp drat
                                    wt qsec vs am gear carb
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3 14.7
```

What to do?

- Using non-standard evaluation functions inside your own functions can cause surprises
- Avoid using non-standard evaluation functions inside your functions
- Or, learn the surprising cases and protect against them





Let's practice!





Hidden arguments

Pure functions

- 1. Their output only depends on their inputs
- 2. They don't affect the outside world except through their return value
- Hidden arguments are function inputs that may be different for different users or sessions
- Common example: argument defaults that depend on global options



Viewing global options

```
> options()
$add.smooth
[1] TRUE
> options()
$add.smooth
[1] TRUE
$browserNLdisabled
[1] FALSE
$CBoundsCheck
[1] FALSE
$check.bounds
[1] FALSE
• • •
```



Getting and setting options

```
> getOption("digits")
[1] 7
> options(digits = 5)
> getOption("digits")
[1] 5
# To read about some of the common options
> ?options
```



Relying on options in your code

- The return value of a function should never depend on a global option
- Side effects may be controlled by global options





Let's practice!





Wrappingup



Writing functions

- If you have copy-and-pasted two times, it's time to write a function
- Solve a simple problem, before writing the function
- A good function is both correct and understandable

Functional Programming

- Abstract away the pattern, so you can focus on the data and actions
- Solve iteration problems more easily
- Have more understandable code

Remove duplication and improve readability

```
> library(purrr)
> df[] <- map(df, rescale01)</pre>
```



Unusual inputs and outputs

- Deal with failure using safely()
- Iterate over two or more arguments
- Iterate functions for their side effects

Write functions that don't surprise

- Use stop() and stopifnot() to fail early
- Avoid using type-inconsistent functions in your own functions
- Avoid non-standard evaluation functions in your own functions
- Never rely on global options for computational details

Wrapping up

- Solve the problem that you're working on
- Never feel bad about using a for loop!
- Get a function that works right, for the easiest 80% of the problem
- In time, you'll learn how to get to 99% with minimal extra effort
- Concise and elegant code is something to strive towards!





Thanks!