# MFE R Programming Workshop Week 2

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## Introduction

#### Questions

Any questions before we start?

#### Overview of Week 2

- Functions
- Optimization
- Control Statements
- ► Lab

#### **Functions**

# Everything that happens in R is a function call

## [1] 15

```
`<-`(mynumber, 3)</pre>
print(mynumber)
## [1] 3
^+ (3,4)
## [1] 7
a <- : (11,20)
[(a,5)]
```

#### **Function Definitions**

```
myfunc <- function(x) x^2
myfunc(10)</pre>
```

```
## [1] 100
```

- The last value evaluated is what is returned by the function.
- ► You can also write return(x^2).
  - This is useful if you want to break out of the function early.

#### Scope Rules for Functions

# You can't access N out here

▶ Variables defined inside a function are local to that function.

```
myfunc <- function(x) {
    N <- 10
    N * x^2 # return(N*x^2) is optional
}
myfunc(10)
## [1] 1000</pre>
```

#### **Default Arguments**

## [1] 2

- ▶ R makes frequent use of default arguments.
- ▶ Default arguments are initialized with a default value unless the user specifies an override.

```
g \leftarrow function(x, y=2, z=2) x*y*z
g(2)
## [1] 8
g(2, 1)
## [1] 4
g(2, 1, 1)
```



## The Pipe Operator %>%

- Pipes are a powerful tool for clearly expressing a sequence of multiple operations.
- The magrittr package provides a pipe operator.
- ► See vignette("magrittr").
- Basic piping:
  - x %>% f is equivalent to f(x)
  - x %>% f(y) is equivalent to f(x, y)
  - x %>% f %>% g %>% h is equivalent to h(g(f(x)))
- The argument placeholder:
  - x %>% f(y, .) is equivalent to f(y, x)
  - x %% f(y, z = .) is equivalent to f(y, z = x)

## An Example of %>%

```
library(magrittr)
x < -1:10
x %>% mean
## [1] 5.5
y \leftarrow c(x, NA)
y %>% mean(na.rm=TRUE)
## [1] 5.5
```

## Expose the variables with %\$%

➤ The %\$% allows variable names (e.g. column names) to be used in a function.

```
iris %>%
  subset(Sepal.Length > mean(Sepal.Length)) %$%
  cor(Sepal.Length, Sepal.Width)
```

```
## [1] 0.3361992
```

## Compound assignment pipe operations with %<>%

► There is also a pipe operator which can be used as shorthand notation in situations where the left-hand side is being "overwritten":

```
iris$Sepal.Length <-
iris$Sepal.Length %>%
sqrt()
```

▶ Use the %<>% operator to avoid the repetition:

```
iris$Sepal.Length %<>% sqrt
```

► This operator works exactly like %>%, except the pipeline assigns the result rather than returning it.

## **Control Statements**

#### **WARNING!**

- ► In general, loops are *slow* in R.
- Vectorized code is faster.
- ▶ Loops are often a good place to write a R function in C++.
  - ► We will cover this later.

#### Looping vs Vectorization

```
library(microbenchmark)
sq <- function(x) {</pre>
 n <- length(x)
  res <- rep(NA, n)
  for(i in 1:n) res[i] <- x[i]^2</pre>
  res
x < -1:1000
microbenchmark(
  sq(x),
 x^2
```

```
## Unit: microseconds
## expr min lq mean median uq max neval
## sq(x) 70.747 71.841 132.54071 74.394 135.294 3738.982 100
## x^2 1.459 1.824 2.80129 2.553 3.283 9.117 100
```

# For loops (1)

- A for loop iterates over an index, provided as a vector.
- ➤ To iterate over the length of a vector x, we can either use 1:length(x) or seq(x).
  - seq(x) protects against zero-length vectors.

```
x <- c(1:5)
y <- NULL # we need to initialize an empty vector
for(i in seq(x)) { # safer than 1:length(x)
    y[i] <- x[i] + 2
}
y</pre>
```

```
## [1] 3 4 5 6 7
```

# For loops (2)

► Another nice way to make a for loop.

```
x <- c(2:4)
for(i in x) {
    print(i + 2)
}

## [1] 4
## [1] 5
## [1] 6</pre>
```

## While loops

► A while loop runs the code inside the braces repeatedly as long as the tested condition proves TRUE.

```
x <- c(1:5)
y <- NULL
i <- 1
while(i <= length(x)) {
   y[i] <- x[i] + 2
   i <- i + 1
}
y</pre>
```

```
## [1] 3 4 5 6 7
```

#### If Statements

- ▶ If statements operate on length-one logical vectors.
- Syntax:

```
if(cond1=true) { cmd1 } else { cmd2 }
```

```
myabs <- function(x) {
   if(x < 0) {
      myabs <- -x
   }
   x
   return(myabs)
}
x <- -10
myabs(x)</pre>
```

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

#### Ifelse Statements

- ▶ Ifelse statements operate on vectors of variable length.
- Syntax:
  - ifelse(test, true\_value, false\_value)

```
x <- 1:10 # Creates sample data
ifelse(x<5 | x>8, x, 0)
## [1] 1 2 3 4 0 0 0 0 9 10
```

# Optimization in ${\sf R}$

#### Overivew

- ▶ R has many tools to solve optimization problems.
- See CRAN Task View: Optimization and Mathematical Programming.
- ► Continuous Global Optimization in R

## One Dimensional Root (Zero) Finding

- ► The function uniroot searches an interval for a root (i.e., zero) of the function f with respect to its first argument.
- Suppose we want x such that  $5 e^{x^2} = x^2$ .
- ▶ We convert this to the root finding problem  $5 e^{x^2} x^2 = 0$ .

```
f <- function(x) {
   5 - exp(x^2) - x^2
}
root <- uniroot(f, c(0, 2), tol = 10^-8)
root$root</pre>
```

```
## [1] 1.143048
```

```
f(root$root)
```

```
## [1] 9.738421e-09
```

#### General-Purpose Optimization with optim

► Rosenbrock Banana function

```
fr <- function(x) {  ## Rosenbrock Banana function
        x1 <- x[1]
        x2 <- x[2]
        100 * (x2 - x1 * x1)^2 + (1 - x1)^2
}
optim(c(-1.2,1), fr)$par</pre>
## [1] 1.000260 1.000506
```

#### Investments Problem Set 2

► You will use R to calculate yield-to-maturity.

## Lab 1

#### Reading in Data for Lab 1

- read.table is the basic function to read in tabular data.
- read.csv is a special case of read.table.
  - ► As usual see ?read.table.
  - Often you want to set stringsAsFactors = FALSE.

```
## S0 sigma r T K
## 1 100 0.3 0.0 1 100
## 2 101 0.3 0.0 1 100
## 3 101 0.1 0.1 1 105
```

- optdata is a data.frame, a specialized type of list
- write.csv writes data to a .csv file.

#### Lab 1

Let's work on Lab 1.