

September, 2018





R is a program. Like any program, it takes inputs from a user and does something.

In this case, the input is your code.

You can input code in two ways:

- 1. R can read it from a file
- 2. R can read it "interactively"

We'll focus on using R interactively in this module.





What do we want R to do? What is the output?

This will determine how you use R (or any other language).





If you want to communicate something to someone:

- HTML
- PDF Document
- Jupyter Notebook
- Website
- Pretty Pictures

Plus the code to repeat/modify the above material





If you want to transform data:

• CSV, JSON, RData, SQL, database entries

Plus the code to repeat/modify the transformation



# **Your Own Opinion**

If you want to learn, ask questions, the output might be:

Nothing

Plus the code to repeat/modify your question or answer.



## You're writing code

No matter what you use R for, the code is always an end in and of itself.



#### R is functional

You can manage code complexity in R with functions.

- 1. Write functions that have one responsibility.
- 2. Assign each responsibility fully to one function.





Today we want to use R to do the following:

Test the accuracy of the standard errors in an OLS regression, given input data of different sizes, with Gaussian noise.



### Break down the problem

The first step is to break down the task into separate responsibilities:

- 1. Generate input data with Gaussian noise, of differing sizes.
- Get standard errors of OLS regression on the above input data.
- 3. Compare OLS estimate to the true value.



### Responsibility

Given our breakdown of the tasks in this problem, we know we need 3 functions.

Let's start with the first one, generating data.





Here is a function that generates zeros:

```
generate_data <- function (howMuchData) {
   output <- rep(0, how_much_data)
   return(output)
}</pre>
```

How does variable assignment work? Function arguments? Function calls?





Remember the return statement is optional. This is the same:

```
generate_data <- function (how_much_data) {
   output <- rep(0, how_much_data)
   output
}</pre>
```



What if we do this? Is this the same function?
how\_much\_data <- 20
generate\_data <- function () {
 output <- rep(0, how\_much\_data)
 output
}</pre>



If R doesn't find a variable inside the function, it will look outside.



```
What's bad about this?
generate_data <- function () {</pre>
    output <- rep(0, how_much_data)</pre>
    output
}
how much data <- 20
data <- generate data()</pre>
how_much_data <- 35
data_2 <- generate_data()</pre>
```





Remember the responsibility metaphor. Previously, the function generate\_data was completely in charge of generating data.

In this version, the responsibility of generating the data is shared: between the function and whoever is in charge of keeping track of the value of "how\_much\_data" in the global scope.





That someone is you.

But you shouldn't be sharing the responsibility. You have enough to do.

Give the responsibility fully to the function. Write a reliable function.





```
generate_data <- function (how_much_data) {
   output <- rep(0, how_much_data)
   output
}
data <- generate_data(20)
data_2 <- generate_data(35)</pre>
```





Let's work on generating data according to the data generating process:

```
y <- beta * x + eps
```

#### Where

```
x <- rnorm(1, mean=0, sd=1)
eps <- rnomr(1, mean=0, sd=sd)</pre>
```





```
generate_data <- function(N, beta, sd) {
    x <- rnorm(N, 0, 1)
    eps <- rnorm(N, 0, sd)
    y <- beta*x + eps
    list(x = x, y = y)
}</pre>
```



# **Running the Regression**

```
run_regression <- function(y, x) {
   coef <- ???
   se <- ???
   list(coef=coef, se=se)
}</pre>
```

Here again we have two responsibilities:

- 1. Calculating the coefficient on x.
- 2. Calculating the standard errors on the coefficients.



# **Running the Regression**

```
calc_coef <- function(y,x) cov(x,y) / var(x)</pre>
calc_se <- function(y, x, coef) {</pre>
    n <- length(y)
    eps \leftarrow y - x*coef
    e sd <- mean(eps^2)
    se <- sqrt(e sd / (n*var(x)))
    se
var() actually returns the unbiased variance estimate (n-1), but
we will just use it to keep things simple.
```



# **Running the Regression**

```
run_regression <- function(y, x) {
   coef <- calc_coef(y, x)
   se <- calc_se(y, x)
   list(coef=coef, se=se)
}</pre>
```



### **Evaluating the Model**

```
eval_model <- function(coef, se, beta, conf = 1.96) {
    up <- coef + se*conf
    down <- coef - se*conf
    beta > down & beta < up
}</pre>
```

Note: default function values, logical operators, returns a scalar or a vector?



# Making our simulation

```
simulate <- function(N, beta, sd) {
    d <- generate_data(N, beta, sd)</pre>
    m <- run_regression(d$y, d$x)</pre>
    eval model(m$coef, m$se, beta)
avg simulations <- function(M, N, beta, sd) {
    inside <- sapply(1:M, function (x) {</pre>
        simulate(N, beta, sd)
    })
    sum(inside) / M
}
```

Note: apply family, summing boolean vector





```
Let's see our results:
```

```
library(ggplot2)

check_N <- function(M, beta, sd) {
    x <- seq(4, 50, 2)
    y <- sapply(x, function(N) {
        avg_simulations(M, N, beta, sd)
    })
    qplot(x, y)
}</pre>
```



### **Results**

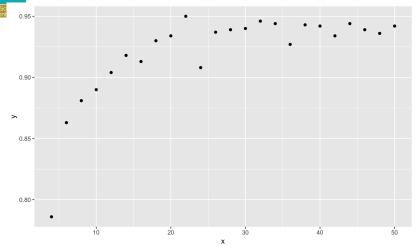


Figure 1: inline