# CSC 121: Computer Science for Statistics

Radford M. Neal, University of Toronto, 2017

http://www.cs.utoronto.ca/~radford/csc121/

Week 10

# Many Ways to Write a Simple Function

In this lecture, we'll look at many ways of writing a simple function called <code>is\_not\_decreasing</code>, which takes one argument, a vector, and returns TRUE if the elements in the vector are in non-decreasing order, and FALSE otherwise. We'll see some new R features along the way.

#### Examples:

```
> is_not_decreasing (c(4,8,8,9))
[1] TRUE
> is_not_decreasing (c(5,1,3))
[1] FALSE
> is_not_decreasing (7)
[1] TRUE
```

We'll assume that the vector has no NA values. What would be a reasonable thing to do if it did?

### Ending a Loop Using a Logical Flag Variable

Here's one solution, that uses the setting of a logical variable as a way of terminating a while loop:

```
is_not_decreasing <- function (v) {</pre>
    answer_is_known <- FALSE</pre>
    i <- 2
    while (!answer_is_known) {
         if (i > length(v)) {
             answer <- TRUE
             answer_is_known <- TRUE</pre>
         else if (v[i] < v[i-1]) {
             answer <- FALSE
             answer_is_known <- TRUE</pre>
         }
         i <- i + 1
    answer
}
```

### Using a repeat Loop and break Statement

This function used two logical variables — one to hold the answer returned, the other to indicate when the answer is now known, and hence the loop can end. We can instead use a loop written using repeat, which continues indefinitely, until a break statement is done:

```
is_not_decreasing <- function (v) {</pre>
    i <- 2
    repeat {
        if (i > length(v)) {
             answer <- TRUE
            break
        }
        if (v[i] < v[i-1]) {
             answer <- FALSE
            break
        i <- i + 1
    }
    answer
}
```

### Using break Within a for Loop

We can use **break** to immediately exit any kind of loop. Here's another way to write this function:

```
is_not_decreasing <- function (v) {
    answer <- TRUE
    if (length(v) > 1)
        for (i in 2:length(v)) {
        if (v[i] < v[i-1]) {
            answer <- FALSE
            break
        }
     }
     answer
}</pre>
```

In this version, we initially set answer to TRUE, which will be the answer if we don't find a place where the elements decrease. If we do find a decrease, we set answer to FALSE, and also immediately exit the for loop.

Caution: The break statement exits from the innermost loop that contains it. If you're inside two loops, you can't use break to exit both of them at once.

#### Returning a Value for a Function Immediately

Rather than exit a loop with break after setting answer, and then making answer the value of the function by putting it as the last thing, we can instead use return to exit the whole function, and specify the value it returns.

At the end, we could just have written TRUE instead of return(TRUE) — they do the same thing at the end of a function.

Why is the check for length(v) > 1 needed?

#### Avoiding Loops with a Vector Comparison

We can write is\_not\_decreasing without an R loop using a vector comparison and the all function:

```
is_not_decreasing <- function (v) all (v[-length(v)] <= v[-1])
```

In this version, v[-length(v)] will contain all of v except the last element, and v[-1] will contain all of v except the first element. So  $v[-length(v)] \le v[-1]$  compares each element except the last to the next element. The vector v is non-decreasing if all these comparisons are TRUE.

Here's another way to do the same thing:

```
is_not_decreasing <- function (v) {
   if (length(v) < 2)
        TRUE
   else
      all (v[1:(length(v)-1)] <= v[2:length(v)])
}</pre>
```

Why is the check for length(v) < 2 needed here, but not in the version above?

#### Recursion — When a Function Calls Itself

As you know, an R function can call another R function, which can call yet another R function, etc.

Indeed, an R function can even call itself. This is called "recursion".

Of course, a function had better not *always* call itself, or it will just keep calling, and calling, and calling, without end.

But having a function sometimes call itself can be useful. Here's a recursive function to compute factorials in R:

fact <- function (n) if (n == 0) 1 else n \* fact(n-1)
(Although R already has a pre-defined factorial function.)</pre>

In fact, anything computable can be computed using if and recursion, without any loops or assignment statements. That's not a typical style of programming in R, but it is typical for some other programming languages.

```
Two Recursive Versions of is_not_decreasing
We could write the is_not_decreasing function using recursion. Here's one way:
    is_not_decreasing <- function (v) {</pre>
         if (length(v) \le 1)
             TRUE
         else if (v[2] < v[1])
             FALSE
         else
             is_not_decreasing(v[-1])
    }
Here's another way that doesn't copy parts of v, and also extends the function's
meaning so it checks only from a certain point forward (default, from the start):
    is_not_decreasing <- function (v, from=1) {</pre>
         if (length(v) <= from)</pre>
             TRUE
         else if (v[from+1] < v[from])</pre>
             FALSE
         else
             is_not_decreasing(v,from+1)
    }
```

#### Operations on Vectors

We've seen before that R can do many operations on entire vectors (or matrices), not just on single numbers. For example, we can add 1 to all elements of a vector:

Instead of the statement  $v \leftarrow u + 1$  we could have written a loop:

```
> v <- u
> for (i in 1:length(v)) v[i] <- v[i] + 1
> v
[1] 4 6 2 10
```

But v <- u + 1 is easier to write, easier to read, and also faster in R.

This isn't magic, though — there still **is** a loop hidden within the implementation of R, and in some other languages writing a loop yourself would be just as fast.

R has many other facilities for doing operations on vectors, matrices, or lists without having to write a loop, which often are also faster.

# Replacing Loops with "apply" Functions

Functions in the "apply" family take as arguments both a data structure and a function to apply to parts of the data structure — an example of "functional programming", using functions to construct more complex operations.

The lapply function operates on a list, and returns a list of results of applying a given function to each element of the list. Here's an example using the is.numeric function, which says whether something is a numeric vector:

```
> L <- list ("abc", c(123,456), TRUE)
> lapply(L,is.numeric)
[[1]]
[1] FALSE

[[2]]
[1] TRUE
[[3]]
[1] FALSE
```

### Using "apply" on Matrices

You can use apply to apply a function to all rows or to all columns of a matrix.

If the function applied returns a single value, the result is a vector of these values:

If the function returns a vector of length greater than one, the result is a matrix:

```
> apply (M, 1, function (v) c(sum(v), prod(v)))
      [,1] [,2]
[1,] 9 12
[2,] 15 48
```

#### Logical Operators

Some previous slides have mentioned logical operations on vectors. These operate on operate on vectors of logical values, returning a vector of logical values.

For one logical value, the operators are defined as follows:

- ! Logical "not": TRUE if its operand is FALSE, FALSE if its operand is TRUE.
- & Logical "and": TRUE only if both operands are TRUE.

TRUE

Logical "or": TRUE if either operand is TRUE.

When applied to logical vectors, the operations are done on each element in turn:

```
> a <- c (TRUE, TRUE, FALSE, FALSE)
> b <- c (TRUE, FALSE, TRUE, FALSE)
> a & b
[1] TRUE FALSE FALSE FALSE
> a | b
[1] TRUE TRUE TRUE FALSE
> !a
```

[1] FALSE FALSE TRUE

# An Example of apply Using Logical Operations

Here's how apply can be used to see which columns in a matrix have values that are all in the range of the first value to the last value: