

# Programming with functions

## Functions

A function has a form:

```
name <- function(argument_1, argument_2, ...) {  
  expression_1  
  expression_2  
  ...  
  return(output)  
}
```

- `argument_1`, `arguments_2` are the names of variables.
- `name` is the name of function. You change `name` as you want.

To call or run the function we type:

```
name(x1, x2, ...)
```

the value of this expression is the value of output.

- A function may have more than one `return` statement, in which case it stops after executing the first one it reaches.
- If there is no statement `'return(output)'` then the value returned by the function is the value of the last expression in the braces.
- A function always returns a value.
- For some functions the value returned is unimportant.
- In such cases one usually omits the return statement, or returns NULL.
- If the value returned by a function is not assigned to a variable, then it is printed.
- The most important advantage of using a function is that once it is loaded, it can be used again and again without having to reload it.
- The second most important use of functions is to break down a programming task into smaller logical units.

**Example : Find zeros of  $a_2x^2+a_1x+a_0=0$**

```
quad3 <- function(a0, a1, a2) {  
  if(a2 == 0 && a1 == 0 && a0 == 0) {  
    roots <- NA  
  } else if (a2 == 0 && a1 == 0) {  
    roots <- NULL  
  } else if ( a2 == 0 ) {  
    roots <- -a0/a1  
  } else {  
    discrim <- a1^2 - 4*a2*a0  
    if (discrim > 0) {  
      roots <- (-a1 +c(1,-1) * sqrt(a1^2-4*a2*a0))/(2*a2)  
    } else if (discrim == 0) {  
      roots <- -a1/(2*a2)  
    } else {  

```

```

    roots <- NULL
  }
}
return(roots)
}

```

### Example : n choose r

The number of ways that you can choose r things from a set of n, ignoring the order, is

$$\frac{n!}{r!(n-r)!}$$

```

n_factorial <-function(n) {
  n_fact <- prod(1:n)
  return(n_fact)
}
n_choose_r <- function(n, r) {
  n_ch_r <- n_factorial(n)/n_factorial(r)/n_factorial(n-r)
  return(n_ch_r)
}

```

### Example : Winsorised mean

k-th Winsorised mean of  $x=\{x_1, \dots, x_n\}$  is defined as

$$w_k = \frac{(k+1)x_{k+1} + x_{k+2} + \dots + x_{n-k+1} + (k+1)x_{n-k}}{n}$$

```

wmean <- function(x, k) {
  x <- sort(x)
  n <- length(x)
  x[1:k] <- x[k+1]
  x[(n-k+1):n] <- x[n-k]
  return(mean(x))
}

```

### Exmple : Swap

swap values of x[1] and x[2]

```

f1 swap <- function(x){
f2   y <- x[2]
f3   x[2] <- x[1]
f4   x[1] <- y
f5   return(x)
f6 }
p1 x <- c(7, 8, 9)
p2 x[1:2] <- swap(x[1:2])
p3 x[2:3] <- swap(x[2:3])

```

## Scope and its consequences

- Argument and variables defined within a function exist only within that function.
- If you define and use a variable `x` inside a function, it does not exist outside the function.
- If variables with the same name exist inside and outside a function, then they are separate and do not interact at all.
- The variable defined outside the function can be seen inside the function (provided there is not a variable with the same name defined inside).

```
test <- function(x) {  
  y <- x+1  
  return(y)  
}  
test(1)  
  
## [1] 2  
y  
### Error : Object "y" not found
```

```
test2 <- function(x) {  
  y <- x + z  
  return(y)  
}  
z <- 1  
test2(1)
```

```
## [1] 2  
z<-2  
test2(1)  
  
## [1] 3
```

## Optional argument and default value

- To give argument\_1 the default value `x1` we use `argument_1 = x1` within the function definition.
- If an argument has a default then it may be omitted when calling the function, in which case the default is used.

```
test3 <- function(x=1, y=1, z=1) {  
  return(x*100+y*10+z)  
}  
test3(2,2)  
  
## [1] 221  
test3(y=2, z=2)  
  
## [1] 122
```

## Vector-based programming

- Many R functions are vectorised.
- To further facilitate vector-based programming, R provides functions that enable the vectorisation of user-defined functions.
- `apply`, `sapply`, `lapply`, `tapply`, and `mapply`.

- `sapply(X, FUN)`
- apply function `FUN` to every element of vector `X`.

### Example : Density of primes – `sapply`

- Let  $\rho(n)$  be the number of primes less than or equal to  $n$
- Then

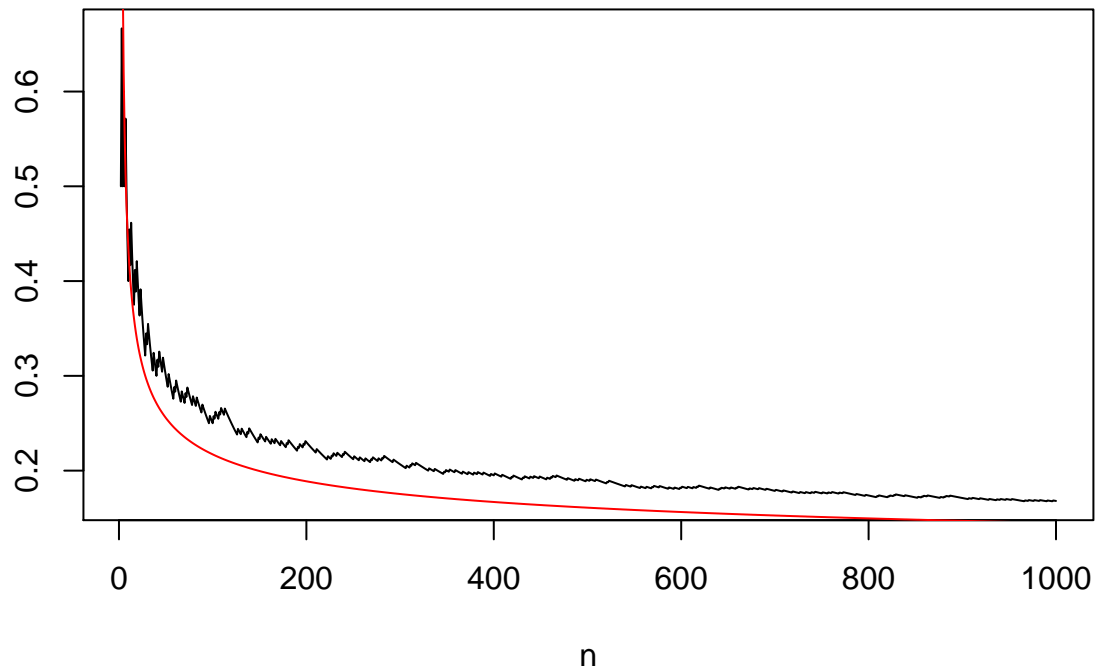
$$\lim_{n \rightarrow \infty} \frac{\rho(n) \log(n)}{n} \rightarrow 1.$$

- To check this, first we define a function `prime` that tests if a given integer is prime.
- We then use `sapply` to apply `prime` to the vector `2:n`.

```
prime <- function(n) {
  if ( n==1 ) {
    is.prime <- FALSE
  } else if ( n==2 ) {
    is.prime <- TRUE
  } else {
    is.prime <- TRUE
    for ( m in 2:(n/2) ) {
      if ( n%%m == 0 ) is.prime <- FALSE
    }
  }
  return(is.prime)
}
```

```
n <- 1000
m.vec <- 2:n
primes <- sapply(m.vec, prime)
num.primes <- cumsum(primes)
plot(m.vec, num.primes/m.vec, type = "l", main = "prime density", xlab = "n", ylab = "")
lines(m.vec, 1/log(m.vec), col= "red")
```

## prime density



### Recursive programming

- A recursive program is one that calls itself.
- This is useful because many algorithms are recursive in nature.

#### n factorial

```
nfact2 <- function(n) {  
  if (n==1) {  
    cat("called nfact2(1)\n")  
    return(1)  
  } else {  
    cat("called nfact2(", n, ")\n", sep="")  
    return(n*nfact2(n-1))  
  }  
}
```

### Sieve of Eratosthenes

The Sieve of Eratosthenes is an algorithm for finding all of the primes less than or equal to a given number  $n$ .

1. Start with the list  $2, 3, \dots, n$  and  $p = 2$ .
2. Remove from the list all elements that are multiples of  $p$  (but keep  $p$  itself).
3. Increase  $p$  to the smallest element of the remaining list that is larger than the current  $p$ .
4. If  $p$  is larger than  $\sqrt{n}$  then stop, otherwise go back to step 2.

```
primesieve <- function(sieved, unsieved) {  
  p <- unsieved[1]  
  n <- unsieved[length(unsieved)]
```

```
if ( p^2 > n ) {  
  return(c(sieved, unsieved))  
} else {  
  unsieved <- unsieved[unsieved %% p != 0]  
  sieved <- c(sieved, p)  
  return(primesieve(sieved, unsieved))  
}  
}
```