# Advanced R programming: solutions 1 Dr Colin Gillespie May 2, 2014

## 1 Argument matching

R allows a variety of ways to match function arguments.<sup>1</sup> We didn't cover argument matching in the lecture, so let's try and figure out the rules from the examples below. First we'll create a little function to help

<sup>1</sup> For example, by position, by complete name, or by partial name.

```
arg_explore = function(arg1, rg2, rg3)
paste("a1, a2, a3 = ", arg1, rg2, rg3)
```

Next we'll create a few examples. Try and predict what's going to happen before calling the functions

One of these examples will raise an error - why?

```
arg_explore(1, 2, 3)
arg_explore(2, 3, arg1 = 1)
arg_explore(2, 3, a = 1)
arg_explore(1, 3, rg = 1)
```

Can you write down a set of rules that R uses when matching arguments?

```
## SOLUTION
## See http://goo.gl/NKsved for the offical document
## To summeriase, matching happens in a three stage pass:
#1. Exact matching on tags
#2. Partial matching on tags.
#3. Positional matching
```

Following on from the above example, can you predict what will happen with

```
plot(type = "l", 1:10, 11:20)
```

and

```
rnorm(mean = 4, 4, n = 5)
```

```
## SOLUTION
#plot(type="l", 1:10, 11:20) is equivilent to
plot(x=1:10, y=11:20, type="l")
#rnorm(mean=4, 4, n=5) is equivilent to
rnorm(n=5, mean=4, sd=4)
```

### The ... argument

A common argument<sup>2</sup> is . . . . We can explore what happens using the print\_dots function in the nclRadvanced package

```
<sup>2</sup> Especially when dealing with S<sub>3</sub> ob-
jects and functions.
```

```
library("nclRadvanced")
arg_explore2 = function(arg1 = 5, ...)
  print_dots(...)
```

- Repeat the examples used in arg\_explore, but include the ... argument.
- Functions as first class objects

Suppose we have a function that performs a statistical analysis

```
## Use regression as an example
stat_ana = function(x, y) {
    lm(y \sim x)
```

However, we want to alter the input data set using different transformations<sup>3</sup>. In particular, we want the ability to pass arbaritary transformation functions to stat\_ana.

<sup>3</sup> For example, the log transformation.

• Add an argument trans to the stat\_ana function. This argument should have a default value of NULL.

```
## SOLUTION
stat_ana = function(x, y, trans=NULL) {
  lm(y \sim x)
```

• Using is.function to test whether a function has been passed to trans, transform the vectors x and y when appropriate. For example,

```
stat_ana(x, y, trans = log)
```

would take log's of x and y.

```
## SOLUTION
stat_ana = function(x, y, trans=NULL) {
  if(is.function(trans)) {
    x = trans(x)
    y = trans(y)
  lm(y \sim x)
```

• Allow the trans argument to take character arguments in additional to function arguments. For example, if we used trans = 'normalise', then we would normalise the data4.

<sup>4</sup> Subtract the mean and divide by the standard deviation.

```
## SOLUTION
stat_ana = function(x, y, trans=NULL) {
  if(is.function(trans)) {
    x = trans(x)
    y = trans(y)
  } else if (trans == "normalise") {
    x = scale(x)
    y = scale(y)
  }
  lm(y \sim x)
```

## Variable scope

Scoping can get tricky. Before running the example code below, predict what is going to happen

1. A simple one to get started

```
f = function(x) return(x + 1)
f(10)
##Nothing strange here. We just get
f(10)
## [1] 11
```

2. A bit more tricky

```
f = function(x)  {
    f = function(x)  {
        x + 1
    }
    x = x + 1
    return(f(x))
}
f(10)
```

3. More complex

```
f = function(x)  {
    f = function(x)  {
        f = function(x) {
            x + 1
```

```
}
    x = x + 1
    return(f(x))
}
    x = x + 1
    return(f(x))
}
f(10)
```

```
## Solution: The easiest way to understand
## is to use print statements
f = function(x)  {
    f = function(x) {
       f = function(x)  {
           message("f1: = ", x)
           x + 1
        }
       message("f2: = ", x)
       x = x + 1
       return(f(x))
    message("f3: = ", x)
   x = x + 1
   return(f(x))
f(10)
## f3: = 10
## f2: = 11
## f1: = 12
## [1] 13
```

```
4. f = function(x) {
    f = function(x) {
        x = 100
        f = function(x) {
            x + 1
        }
        x = x + 1
        return(f(x))
    }
    x = x + 1
    return(f(x))
}
```

```
## Solution: The easiest way to understand
## is to use print statements as above
```

#### Function closures

Following the examples in the notes, where we created a function closure for the normal and uniform distributions. Create a similar closure for

• the Poisson distribution,<sup>5</sup>

```
<sup>5</sup> Hint: see rpois and dpois.
```

```
poisson = function(lambda) {
    r = function(n = 1) rpois(n, lambda)
    d = function(x, log = FALSE) dpois(x,
       lambda, log = log)
    return(list(r = r, d = d))
```

• and the Geometric distribution.<sup>6</sup>

```
<sup>6</sup> Hint: see rgeom and dgeom.
```

```
geometric = function(prob) {
    r = function(n = 1) rgeom(n, prob)
    d = function(x, log = FALSE) dgeom(x,
        prob, log = log)
    return(list(r = r, d = d))
```

#### 6 Mutable states

In chapter 2, we created a random number generator where the state, was stored between function calls.

- Reproduce the randu generator from the notes and make sure that it works as advertised.
- When we initialise the random number generator, the very first state is called the seed. Store this variable and create a new function called get\_seed that will return the initial seed, i.e.

```
r = randu(10)
r$r()
## [1] 0.0003052
r$get_state()
## [1] 655390
r$get_seed()
## [1] 10
```

```
## Solutions - see below
```

• Create a variable that stores the number of times the generator has been called. You should be able to access this variable with the function get\_num\_calls

```
r = randu(10)
r$get_num_calls()
## [1] 0
r$r()
## [1] 0.0003052
r$r()
## [1] 0.001831
r$get_num_calls()
## [1] 2
```

```
## Solutions
randu = function(seed) {
   state = seed
    calls = 0 #Store the number of calls
    r = function() {
        state <<- (65539 * state)%2^31
        ## Update the variable outside of this
        ## enviroment
        calls <<- calls + 1
        state/2<sup>31</sup>
    }
    set_state = function(initial) state <<- initial</pre>
    get_state = function() state
    get_seed = function() seed
    get_num_calls = function() calls
    list(r = r, set_state = set_state, get_state = get_state,
        get_seed = get_seed, get_num_calls = get_num_calls)
r = randu(10)
r$r()
## [1] 0.0003052
r$get_state()
## [1] 655390
```

```
r$get_seed()
## [1] 10
```

## Solutions

Solutions are contained within the course package

```
library("nclRadvanced")
vignette("solutions1", package = "nclRadvanced")
```