Practical 2

Jumping Rivers

Practical 1

Advanced R programming: Practical 2

Rprofile and Renviron

1. Create an .Rprofile file. An easy way of creating the file is to use the R function file.create, so

1. Try adding a few functions to your .Rprofile. See chapter 2 in the notes for help. Use the hidden environment trick. Also take a look at this stackoverflow question

for ideas.

rm(r)

2. Create an .Renviron file and add the path to your packages.

Argument matching

R allows a variety of ways to match function arguments. For example, by position, by complete name, or by partial name. 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

```
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)
```

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 ~ x)
}
```

However, we want to alter the input data set using different transformations. For example, the log transformation. In particular, we want the ability to pass arbitrary transformation functions to stat_ana.

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 ~ x)
```

• Use 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 ~ 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 data i.e. 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 ~ 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. A bit more tricky
f = function(x) {
    f = function(x) {
        x + 1
    }
    x = x + 1
    return(f(x))
}
f(10)
1. 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.
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))
}
f(10)
## 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>1</sup>

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>2</sup>

geometric = function(prob) {
    r = function(n = 1) rgeom(n, prob)
```

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$r()
r$get_state()
r$get_seed()
```

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$get num calls()
r$r()
r$r()
r$get_num_calls()
## 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
```