# CT561: Systems Modelling and Simulation

# Week 7: Functions, Data Frames and deSolve

https://github.com/JimDuggan/CT561

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#### **Functions**

- A function is a group of instructions that:
  - takes input,
  - uses the input to compute other value, and
  - returns a result (Matloff 2009).
- Users of R should adopt the habit of creating simple functions which will make their work more effective and also more trustworthy (Chambers 2008).
- Functions are declared:
  - using the function reserved word
  - are objects

### Example

- Function declared as an object
- Takes arguments
- Local scope for function variables
- Generates result
- return() explicit
- Implicit last expression evaluated.

```
convC2F<-function(celsius)
{
  fahr<-celsius*9/5 + 32.0
  return(fahr)
}</pre>
```

# Nested Functions: Variable Scope

*f*3

```
a
       10
f4
  b
         100
  ans
         110
res
      110
```

```
f3<-function(a){
   f4<-function(b){
     ans<-a+b
   res < -f4(a^2)
> f3(10)
> r < -f3(10)
[1] 110
```

### Function objects and arguments

- Functions are first-class objects, can be passed to other functions
- Arguments can be named directly in the call
- In that scenario, order of arguments not an issue

```
convert<-function(func, input)
{
   ans<-func(input)
   return (ans)
}</pre>
```

```
> convert(func=convC2F,input=100)
[1] 212
> convert(input=100,func=convC2F)
[1] 212
```

### Challenge 7.1

- Write a function that takes in a vector and returns the number of odd numbers
- Write a function that takes in a vector and returns a vector of even numbers
- Write a function that returns the unique values in a vector

duplicated() determines which elements of a vector or data frame are duplicates of elements with smaller subscripts, and returns a logical vector indicating which elements (rows) are duplicates.

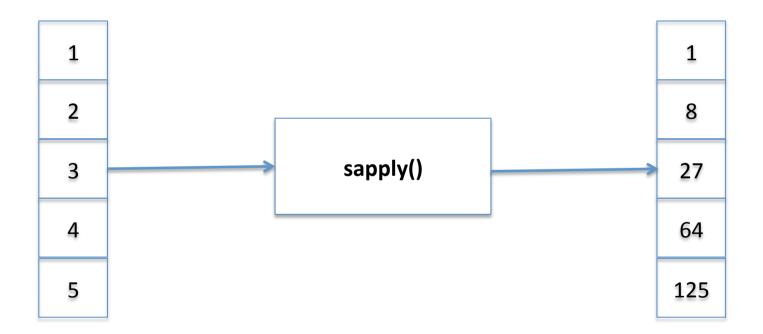
```
> duplicated(c(1,2,3,4,5,6,1))
[1] FALSE FALSE FALSE FALSE FALSE TRUE
```

### sapply()

- Another use of user-defined functions in R is as parameters to the *apply* family of functions, which are one of the most famous and used features of R (Matloff 2009).
- The general form of the sapply(x,f,fargs) function is as follows:
  - x is the target vector or list
  - f is the function to be called and applied to each element
  - fargs are the optional set of arguments that can be applied to the function f.

## Sample Problem

To get the cube of numbers in a vector



#### Method

Embed function in call

### Challenge 7.2

- Use sapply() for the following task
  - transform a vector according to the following function
    - $f(x) = 3x^2 + 5x + 10$

#### **Data Frames**

- A data frame has a twodimensional rows and columns structure.
- Can be viewed as a set of vectors, organised into a column format

```
> ids<-c("1234567","1234568")</pre>
> fNames<-c("Jane","Matt")</pre>
> sNames<-c("Smith","Johnson")</pre>
> ages < -c(21,25)
> ids
[1] "1234567" "1234568"
> fNames
[1] "Jane" "Matt"
> sNames
[1] "Smith" "Johnson"
> ages
[1] 21 25
```

### Creating a Data Frame

```
s<-data.frame(ID=ids,FirstName=fNames,Surname=sNames,</pre>
             Age=ages, stringsAsFactors=FALSE)
       > S
             ID FirstName Surname Age
                     Jane Smith 21
       1 1234567
       2 1234568
                     Matt Johnson 25
       > str(s)
       'data.frame': 2 obs. of 4 variables:
        $ ID : chr "1234567" "1234568"
        $ FirstName: chr "Jane" "Matt"
        $ Surname : chr "Smith" "Johnson"
        $ Age : num 21 25
```

### **Accessing Row Data**

```
> S
      ID FirstName Surname Age
1 1234567
                    Smith 21
             Jane
2 1234568
             Matt Johnson 25
> s[1,]
      ID FirstName Surname Age
                    Smith 21
1 1234567
              Jane
> s[1:2,]
      ID FirstName Surname Age
             Jane Smith 21
1 1234567
2 1234568
             Matt Johnson 25
> s[-1,]
      ID FirstName Surname Age
2 1234568
             Matt Johnson 25
```

### **Accessing Column Data**

```
> S
      ID FirstName Surname Age
              Jane Smith 21
1 1234567
2 1234568 Matt Johnson 25
> s[,"Age"]
[1] 21 25
> s[,1]
[1] "1234567" "1234568"
> s[,2]
[1] "Jane" "Matt"
> s[,3]
[1] "Smith" "Johnson"
> s[,4]
[1] 21 25
```

### Accessing Column Data (using tags)

```
> S
      ID FirstName Surname Age
1 1234567 Jane Smith 21
2 1234568 Matt Johnson 25
> s$ID
[1] "1234567" "1234568"
> s$FirstName
[1] "Jane" "Matt"
> s$Age[1:2]
[1] 21 25
> max(s$Age)
```

### subset() function

```
ID FirstName Surname Age
1 1234567    Jane Smith 21
2 1234568    Matt Johnson 25
> sub<-subset(s,s$Age>21)
> sub
ID FirstName Surname Age
2 1234568    Matt Johnson 25
```

# Adding extra information to a data frame - Vectorized

```
> S
      ID FirstName Surname Age
              Jane Smith 21
1 1234567
2 1234568
             Matt Johnson 25
> s$Discount<-ifelse(s$Age<=21,0.25,0.10)</pre>
> S
      ID FirstName Surname Age Discount
                    Smith 21
              Jane
                                 0.25
1 1234567
                                 0.10
2 1234568
             Matt Johnson 25
```

### Merging Data Frames

- For data analytics, opportunities often arise by merging different data sets into a single data frame, and the merge() function facilitates this
- In our student example, we could have a second data frame that stores examination results.

### Merge Function

```
> S
       ID FirstName Surname Age Discount
                    Smith 21
1 1234567
               Jane
                                    0.25
                                    0.10
2 1234568
               Matt Johnson 25
> res
       ID Subject Grade
1 1234567
            CT111
                     80
2 1234568
            CT111
                     80
> new<-merge(s,res,by="ID")</pre>
> new
       ID FirstName Surname Age Discount Subject Grade
                      Smith
1 1234567
               Jane
                            21
                                    0.25
                                            CT111
                                                     80
             Matt Johnson 25
2 1234568
                                    0.10
                                                     80
                                            CT111
```

### deSolve package



### Journal of Statistical Software

February 2010, Volume 33, Issue 9.

http://www.jstatsoft.org/

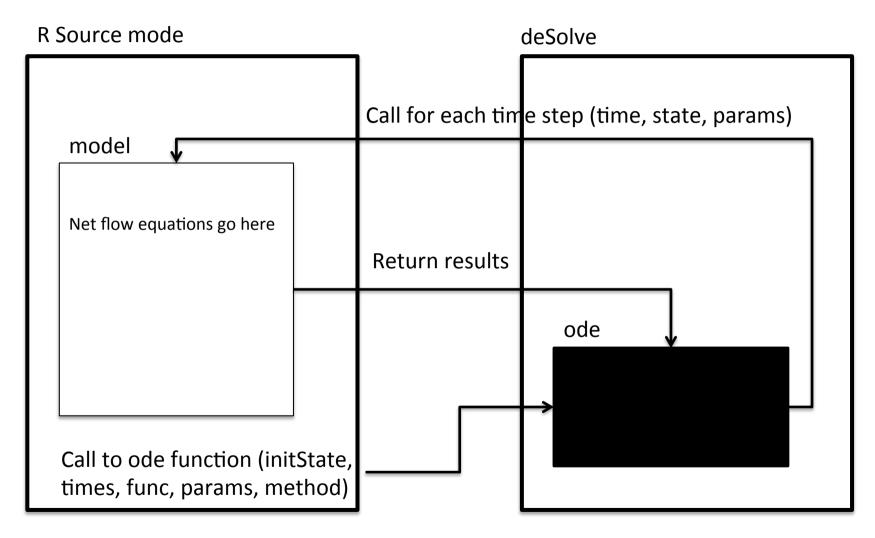
#### Solving Differential Equations in R: Package deSolve

Karline Soetaert
Netherlands Institute of
Ecology

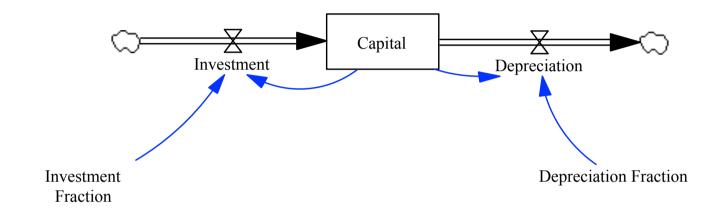
Thomas Petzoldt Technische Universität Dresden

R. Woodrow Setzer US Environmental Protection Agency

### Overall Idea



### Capital Model



Capital= INTEG (Investment-Depreciation, 10000)

Investment Fraction = 0.08

Investment=Capital\*Investment Fraction

Depreciation Fraction=0.05

Depreciation=Capital\*Depreciation Fraction

### R Implementation

```
library(deSolve) # include the deSolve library for odes
library(ggplot2) # include ggplot for displaying graphs

# setup simulation start and finish times, and time step
START<-2015; FINISH<-2035; STEP<-0.25;

# create a simulation time vector, needed by deSolve
simtime <- seq(START, FINISH, by=STEP)</pre>
```

```
> head(simtime)
[1] 2015.00 2015.25 2015.50 2015.75 2016.00 2016.25
> tail(simtime)
[1] 2033.75 2034.00 2034.25 2034.50 2034.75 2035.00
```

### The model function (1/2)

```
10
    # create model function that contains all the equations
11
    # this function receives
12 #
        (1) the current time (time)
# (2) a vectors of stocks (stocks)
14
    # (3) the auxiliary parameters (auxs)
15
16
    model <- function(time, stocks, auxs)</pre>
17 ▼ {
      # with(as.list) makes it easy to reference variable
18
19 -
      with(as.list(c(stocks, auxs)),{
20
        # Calculate the flows
        fInvestment <- sCapital * aInvestmentFraction
21
22
        fDepreciation <- sCapital * aDepreciationFraction
23
```

### The model function (2/2)

```
24
        # Calculate the net flow
        dC_dt <- fInvestment - fDepreciation
25
26
27
        # return calculations as a list,
        # first value is a vector of net flows
28
        return (list(c(dC_dt),
29
30
                      Investment=fInvestment,
                      Depreciation=fDepreciation,
31
                      InvFraction=aInvestmentFraction,
32
                      DepFraction=aDepreciationFraction))
33
34
      })
35
```

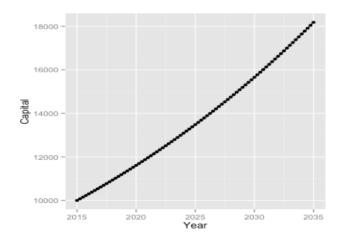
# Create stocks and auxiliaries. Call solver, wrap in data frame

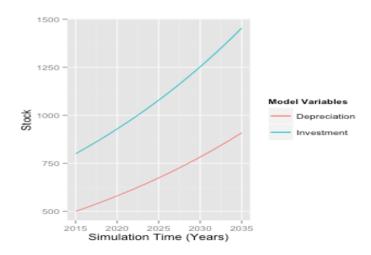
```
37
    # create the vector of stocks and their initial values
38
    stocks <- c(sCapital=10000)
39
40
    # create the vector of auxiliaries
41
    auxs <- c(aInvestmentFraction=0.08,</pre>
42
                  aDepreciationFraction=0.05)
43
    # call the function ode, results returned as a data frame
44
    o<-data.frame(ode(y=stocks, simtime, func = model,
45
46
                       parms=auxs, method="euler"))
17
```

### Simulation Results (in data frame)

```
> str(o)
'data.frame': 81 obs. of 6 variables:
$ time
                 2015 2015 2016 2016 2016 . . .
         : num
$ sCapital : num 10000 10075 10151 10227 10303 ...
$ Investment : num 800 806 812 818 824 ...
$ Depreciation: num 500 504 508 511 515 ...
> head(o)
    time sCapital Investment Depreciation InvFraction DepFraction
1 2015.00 10000.00
                800.0000
                           500,0000
                                        0.08
                                                 0.05
2 2015.25 10075.00
                806,0000
                           503.7500
                                        0.08
                                                 0.05
3 2015.50 10150.56
                812.0450
                           507.5281
                                       0.08
                                                 0.05
4 2015.75 10226.69
                818.1353
                           511.3346
                                       0.08
                                                 0.05
5 2016.00 10303.39
                824.2714
                           515.1696
                                       0.08
                                                 0.05
6 2016.25 10380.67
                           519.0334
                                                 0.05
                830.4534
                                        0.08
```

### Plot results





### Challenge 7.3

- Build an R model for the stock adjustment structure
- Assume S<sub>INIT</sub>=0
- Assume S\* starts at 100 and increases to 200 after 10 time units
- Let AT=1
- Run the simulation for 20 time units with DT=0.125

