

# Running the CoevoMutAntag simulations

*Lucas Arantes Camacho*

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## Introduction

This is a guide to show step-by-step how to run the coevolution simulations in mutualistic networks with cheater exploitation outcomes inserted. You will see that the process uses different functions from the *R/functions* folder depending on what we want to do in our scripts (which are saved in the *R/scripts* folder). Also, you will see that the scripts' base structure has an organization pattern which helps in the understanding of the code steps in the specific function that is being realized.

## Basic structure

We start setting our work directory that must contain the *.txt* file with the empirical matrices that you are interested in working with. Also, we will load some packages and functions that we will need in our script. Remember to install the required packages before running the script.

```
# loading packages and functions
setwd("~/Dropbox/Master/Code/coevo_mut_antag/R/scripts/")

library(ggplot2)
library(reshape2)
library(cowplot)

source("~/Dropbox/Master/Code/coevo_mut_antag/R/functions/Antagonize.R")
source("~/Dropbox/Master/Code/coevo_mut_antag/R/functions/Counting.R")
source("~/Dropbox/Master/Code/coevo_mut_antag/R/functions/FindInteractors.R")
source("~/Dropbox/Master/Code/coevo_mut_antag/R/functions/SpDegree.R")
source("~/Dropbox/Master/Code/coevo_mut_antag/R/functions/CoevoMutAntNet.R")
```

Ok, we have our packages and functions. Now we can think about the matrix we want to work with. In this example I will construct a fully connected bipartite adjacency matrix. If you want to run the scripts with an empirical data matrix just read the matrix with *read.table* function and remember to use the *Square.Matrix* (provided in the *R/functions* folder) if this is a rectangular matrix.

Let's first define our *antprob* value ( $p$ ) which is the probability of the elements from our mutualistic matrix  $M$  pass from mutualism to cheater exploitation. Also, let's construct our  $M$  matrix defining the number of species (animals + plants) that we want in our matrix. It's important to note that we are not interested in intraspecific interactions so diagonal of  $M$  is zeroed.

```
# initial parameters
antprob = 0.8 # current probability value
n_sp = 10 # defining number of species
M = matrix(1, ncol = n_sp, nrow = n_sp) # building matrix M of positive outcomes
diag(M) = 0 # no intraespecific interactions
```

We already have our adjacency matrix  $M$  of positive effects. Now let's apply our first function in the  $M$  matrix: Antagonize. This function creates the  $V$  matrix and transform positive in negative effects follow the antprob value. The functions return the  $M$  and  $V$  matrices as a list which is used to define  $M$  and  $V$  to the next step.

```
# Antagonize M (transform positive links in negative)
antagonize = Antagonize(M, antprob)
M = antagonize[[1]]
V = antagonize[[2]]
```

We have the two layers of our initial matrix, the positive effects layer matrix  $M$  and negative effects layer matrix  $V$ . With those matrices in hand we can apply some functions to get some informations about the numbers of different outcomes interactions in the community and how much each species are interaction with mutualists or cheaters. Let's apply three functions: Counting, FindInteractors and SpDegree. The Counting functions first:

```
c = Counting(M, V)
c

## [[1]]
## [1] 0
##
## [[2]]
## [1] 40
##
## [[3]]
## [1] 5
```

The Counting function provides the counting of the number of interaction outcomes in our adjacency matrix after the Antagonize process. You see a list with three values: [1] number of double negative effects which are not the focus of this work and never reproduced by the Antagonize function, [2] number of cheater exploitation outcomes and [3] number of mutualism outcomes. Useful to do some consistency tests about the frequency of interaction outcomes in the network. Next function, the FindInteractors:

```
f = FindInteractors(M, V)
f

## [[1]]
## integer(0)
##
```

```
## [[2]]
## [1] 1 2 3 4 5 6 7 8 9 10
##
## [[3]]
## [1] 1 2 7 8 9 10
```

The FindInteractors function provides a list with the identity of species that has mutualism and/or cheater exploitation in the community. Due to the way Antagonize works, a single species could be in the cheaters group [[2]] and mutualism group [[3]] at the same time. Useful when we want to separate species in groups depending on the interaction outcome they have. Finally, the last function before we run the coevolution process, the SpDegree:

```
sp = SpDegree(M, V)
sp
```

```
##      AA AM MM
## 1    0  7  2
## 2    0  8  1
## 3    0  9  0
## 4    0  9  0
## 5    0  9  0
## 6    0  9  0
## 7    0  8  1
## 8    0  6  3
## 9    0  8  1
## 10   0  7  2
```

The SpDegree function just counts the interactions of each species separating by outcomes. With that we could know how much each species has of double negative (AA, but not considered in this paper), AM as cheaters exploitation and MM as mutualisms. Useful to identify asymmetries of outcomes between species.

Ok, until now we just explore our matrices of interaction outcomes. Let's work with coevolution now. First, let's define our coevolution model parameters

```
# coevolutionary model parameters
phi = 0.2
alpha = 0.2
theta = runif(n_sp, 0, 5)
init = runif(n_sp, 0, 5)
p = 0.1
epsilon = 3
eq_dif = 0.0001
t_max = 1000
```

After that we gonna use the last (and maybe most important) function that we load in the beginning of this tutorial: the CoevoMutAntNet which runs the coevolution process in the community. You can understand the functioning of this function in the file Guide\_CoevoMutAntag

from my GitHub account.

```
# running coevolution simulation
```

```
z_mat = CoevoMutAntNet(n_sp, M, V, phi, alpha, theta, init, p, epsilon, eq_dif, t_max)
```

The CoevoMutAntNet function provides an final matrix here called z\_mat which has the species in the collums and the timesteps in the rows which each element of this matrix shows the trait values of species. With this matrix we can plot the species traits in time using the reshape2, ggplot2 and cowplot packages to help us. This is the end of the script.

```
# building data frame to plot the results
```

```
traits = as.data.frame(z_mat)
```

```
n_sp = ncol(traits)
```

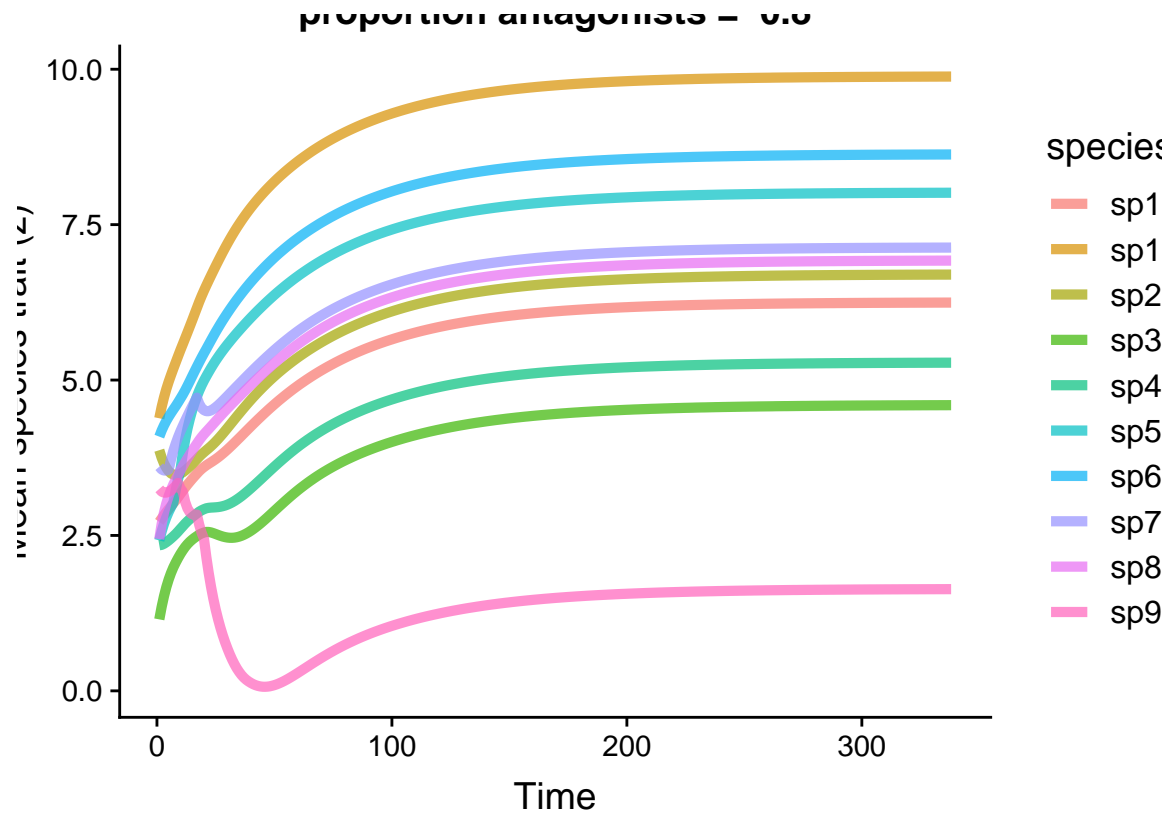
```
traits_vec = c(as.matrix(traits))
```

```
traits_df = data.frame(species = rep(paste("sp", 1:n_sp, sep = ""), each = nrow(traits)),  
                       time = rep(1:nrow(traits), times = n_sp),  
                       trait = traits_vec)
```

```
# plotting traits through time
```

```
plotar = ggplot(traits_df, aes(x = time, y = trait, color = species)) +  
  geom_path(size = 1.8, alpha = 0.7) +  
  ggtitle(paste("proportion antagonists = ", antprob)) +  
  xlab("Time") +  
  ylab("Mean species trait (z)") +  
  theme(axis.text.x = element_text(size = 11),  
        axis.text.y = element_text(size = 11),  
        axis.title = element_text(size = 14),  
        legend.key.size = unit(0.6, "cm"),  
        legend.text = element_text(size = 12))
```

```
plotar
```



## Final considerations

You will see that the scripts used in this paper are just variations of this script-base structure that I show in this small tutorial. For example you could use the `CentralAntagonize` instead of `Antagonize` to transform only the central mutualistic species in cheaters or use some of the `ConDep` functions to considerer the temporal variatio of species outcomes and their influence on coevolution.

Finally, if you have some doubts or suggestions please let me know in [lucas.camacho@usp.br](mailto:lucas.camacho@usp.br).

Good bye.