Simulation Experiments in R

Helene Wagner, University of Toronto

Goals:

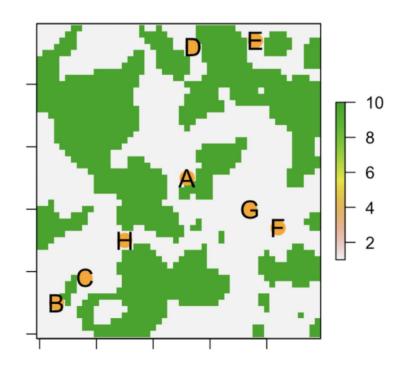
- Simulate a metapopulation on a resistance landscape
- Compare performance of partial Mantel test and Sunder

Methodological Challenges:

Video 1:

- Workflow of a simulation experiment
- Testing statistical methods with simulations
- Partial Mantel test vs. Sunder

Video 2: Efficient R



Simulation Workflow

1. Initialize

Landscape map constant

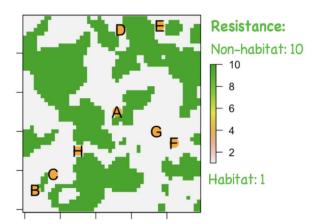
co

Populations (A – H)

constant

Individuals (genotypes)

variable



Create random maps with 'secr':

- Habitat amount (A)
- Habitat aggregation (p)

2. Time step

- Demographic model
- Mating and reproduction
- Dispersal and recruitment

```
> str(para)
List of 15
$ n.pops : num 8
$ n.ind : num 100
$ sex.ratio : num 0.5
$ n.cov : num 3
$ n.offspring: num 2
$ mig.rate : num 0.1
$ disp.max : num 50
$ disp.rate : num 0.05
$ n.allels : num 10
$ n.loci : num 20
$ mut.rate : num 0.001
```

3. Run a single simulation

- Initialize genotypes
- Run for many time steps
- Collect genotype data
- Summarize results

Fst: degree of differentiation Decide: IBD or IBR?

4. Batch run simulations

- Replicate runs with same parameters
- Run scenarios across parameter space
- Store results and settings

		ana	cnaca
Parameter space:	<pre>> para.space rep time</pre>		
	r	ep '	cime
time: # generations	1	1	5
	2	2	5
 rep: # replicate sims 	3	3	5
	4	1	25
	5	2	25
	6	3	25
	7	1	45
	8	2	45
	a	2	15

5. Synthesize results

- Extract summary data
- Visualize in parameter space
- Sensitivity analysis

Robust vs. sensitive



Partial Mantel Tests

IBD IBR PopGenReport::wassermann(eucl.mat = eucl.mat, cost.mats = list(cost=cost.mat), gen.mat = gen.mat, plot=F)\$mantel.tab model <chr> <chr> <chr> IBR | IBD Gen ~cost | Euclidean 0.5366 0.041 IBD | IBR Gen ~Euclidean | cost -0.47530.983

Some issues with (partial) Mantel tests:

- Low statistical power?
- Inflated type I error rates if spatial autocorrelation?

Use simulations to test and compare methods!

Alternative with 'Sunder'

'Bedassle' (Bradburd et al. 2013), alternative implementation in 'Sunder' (Botta et al. 2014)

Run the analysis (parameter settings: http://www.nbi.dk/~botta/Sunder.html#overview)

IBD IBR G+E G E -9050.244 -9058.499 -8974.353

```
> names(which.max(output$mod.lik))
[1] "E"
```

Likelihood

Testing Method Performance

Assessing error rates requires MANY replicate samples!

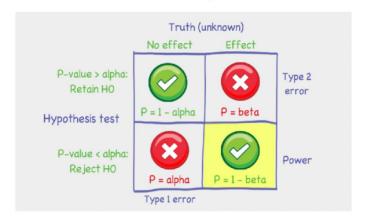
Resistance values

1	1	1	1	1
1	В	1	1	1
1	1	1	1	1
1	1	1	1	1
A	1	1	1	C

Type I error rate

- Simulate under null hypothesis
- Expect alpha % false positives

Statistical Power

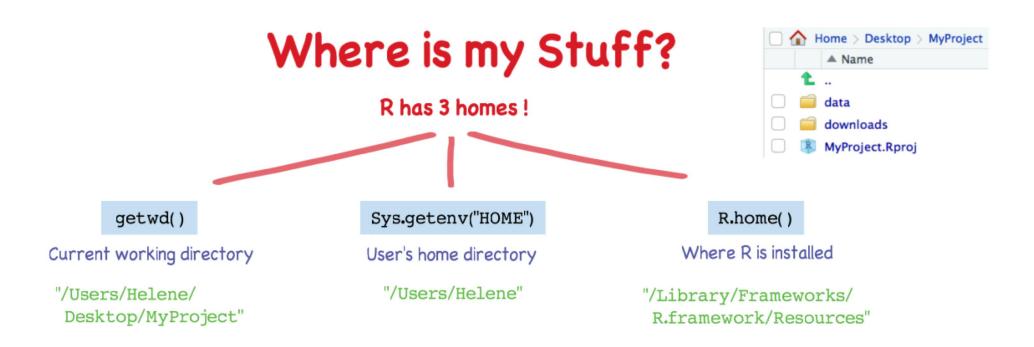


Statistical power to detect effect

- Simulate under alternative hypothesis
- Assess True Positive Rate (TPR)
- Larger effect size -> higher power
- Larger sample size -> higher power

Compare power between methods!





Default working directory?

- Console in regular R session: Varies! Use 'setwd()'

- Console in R project: Project folder

- R Notebook: Notebook file location

Advice



- Always work in an R project
- Use path names relative to project folder
- Use 'here :: here()' in R Notebooks

file.path(here:here(), "data/myFile.csv")

"/Users/Helene/Desktop/MyProject/data/myFile.csv"

"./data/myFile.csv"

"/Users/Helene/Desktop/MyProject"

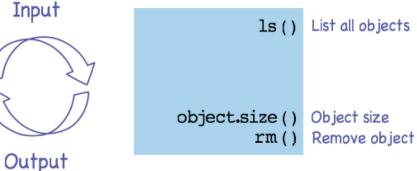
[&]quot;/Users/Helene/Desktop/MyProject/downloads"

Input / Output

File System

```
List all files
                  dir()
  Create directory
                  dir.create()
Download from web
                  download.file()
     Unzip archive
                  unzip()
                  file.size()
          File size
       Remove file
                  file.remove()
```

R Workspace



Text files: csv

```
base ::
                read.table()
                               write.table()
      base ::
                  read.csv()
                               write.csv()
     readr ::
                 read csv()
                               write csv()
data.table ::
                     fread()
                               fwrite()
```

load() save() base :: .RData base :: readRDS() saveRDS() .rds feather: read feather() write feather() .feather

Binary files

```
Package 'rio':
Fast import / export for any file type
```

```
import()
export()
```

rds: Small, fast, flexible object format feather: Compatibility with Python

Warning: check handling of text (character or factor?) and missing values!

Why is my Code Slow?

R was not designed to be fast!

1. Identify bottlenecks

Simple:

- Knit, monitor R Markdown pane
- Name each chunk in R Notebook
- Which chunks take a long time?

Advanced: profiling

- Convert .Rmd to .R: 'purl'
- Source script with 'source'
- Profile with 'lineprof' or 'profvis'
- Visualise time, memory use

Code 113850 **▼** source ▼ withVisible 113840 113840 113840 ► PopGenReport::run.popgensim Week8_vignette_A.. 73970 19060 ▶ getSunder Week8_vignette_A.. ► PopGenReport::wassermann Week8_vignette_A. 7910 mmod::pairwise_Gst_Nei Week8_vignette_A.. 3540 ▶ secr::make.grid

2. Use faster functions

Simple:

- Vectorized: 'lapply' > 'for' loop
- Integrated: tidyverse > R base
- Optimized: CRAN task views

Advanced: benchmarking

- Package 'microbenchmark'
- Define each method as a function
- Compare speed: 'microbenchmark'
- Differences in precision, behavior?

Unit: milliseconds

	and a	2 -		madi au
expr		lq		mediar
import("gen.RData")				
import("gen.rds")				
import("gen.feather")	45.26174	45.96569	46.91915	46.63800
import("gen.csv")	73.59831	80.19661	90.62875	81.40790

3. Speed up your code

Simple:

- Preallocate result vectors
- Don't duplicate large objects
- Use binary data files

Advanced:

- Use 'data.table', 'bigmemory'
- Compile functions: 'cmpfun'
- Parallelize: use multiple cores
- Distribute: cluster computing

mclapply()

foreach()







Bash R scripts 101

myBashFile.sh

Navigate the shell

- Default: same as 'Sys.genenv ("HOME")'
- In RStudio: project folder
- List folder content: 'ls'
- Move to folder with relative path:'cd ./myFolder /subFolder '
- Move up one level: 'cd ...'

Execute a Bash R script

- Must change file permission
- Execute file
- Specify arguments

```
chmod +x myBashFile.sh
./myBashFile.sh 5 0 1
```

Write a Bash R script ...

```
#!/bin/bash
R --slave << EOF

# Your R code:
myFunction <- function (n, m, s)
{
    rnorm (n, m, s)
}
args <- c(5, 0, 1)
myFunction (args)

EOF</pre>
```

knitr :: purl("myNotebook.Rmd")
creates file: myNotebook.R

... with arguments!

```
#!/bin/bash
R --slave --args $@ << EOF

# Your R code:
myFunction <- function (n, m, s)
{
    rnorm (n, m, s)
}
args <- as.numeric (commandArgs ())
myFunction (args)

EOF</pre>
```

Remember what arguments to provide!
Arguments read as 'character' by default.

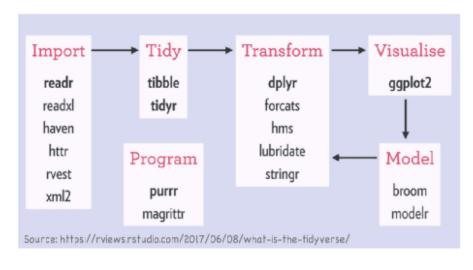
Further Reading

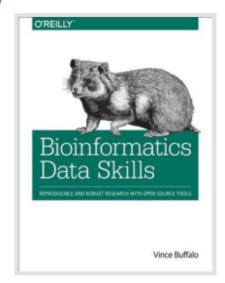
Books

Efficient R programming (Gillespie): https://csgillespie.github.io/efficientR

Advanced R (Wickham): http://adv-r.had.co.nz

R for Data Science (Wickham): http://r4ds.had.co.nz





Tidyverse

- Coherent system of packages for data manipulation, exploration and visualization
- Make data scientists more productive:
 workflow, communication, reproducible research

Blogs

https://www.r-bloggers.com/faster-higher-stonger-a-guide-to-speeding-up-r-code-for-busy-people/https://www.r-bloggers.com/r-with-parallel-computing-from-user-perspectives/https://datascienceplus.com/strategies-to-speedup-r-code/https://support.rstudio.com/hc/en-us/articles/218221837-Profiling-with-RStudiohttps://research.computing.yale.edu/sites/default/files/files/efficientR.pdf

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- Where is my stuff?
- Why is my code slow?

A G F - 2

Interactive R tutorial: generating data, string manipulation

Worked example: landscape genetic simulation experiment

Bonus material: file manipulation, benchmarking, Bash R script