

5085c/6015c Introduction to Landscape Ecology & GIS  
*Spring 2019*



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# Syllabus

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## Introduction of the TAs

- Andres Hernandez-Camacho M.Sc. <https://github.com/andher1802>
- Esteban Correa-Agudelo M.Sc. <https://github.com/maurosc3ner>
- Format of the Lab sessions:
  - This is not a class on statistics !!!!
  - Only a brief introduction about the use of the tool for a better understanding of the landscape ecology applications.
  - Notebooks with theory, practices, code samples, etc.
  - Videos with detailed explanation about the topics.
  - Lab assignment weekly starting on week 2.
- Lab sessions tutorials and materials can be found on <https://maurosc3ner.github.io/6085cintrolandscapeecology/>.

## Content

The goal of this section is to exemplify the role of R in analysis of spatial data in landscape ecology applications. At the beginning, we will review the fundamentals of the use of R within the RStudio interface, basic data structures and operations. Then, we will explore the capabilities of plotting information with R, and finally the use of the library leaflet integrated with RStudio for visualizing spatial data. This brief introduction will help us in understanding the second part of the course, when we will analyze landscape ecology problems and using advance scripts to compute and simulate interesting scenarios of landscape disturbances.

## Schedule

## References

1. Landscape ecology in theory and practice: Pattern and process, second edition (Turner and Gardner, 2015)

Table 1: Lab schedule for the course (changes can take place due to unexpected events)

x
**Week 1 - No assignment**
Introduction to R programming language – Getting familiar with R Studio interface, types of data structures, basic operations
**Week 2 - Lab assignment 2**
Introduction to R programming language - Loops, functions and graphics in R.
**Week 3 – Lab assignment 3**
Introduction to R programming language – Basic statistical operations in R. Proportion test and two sample test, linear regression
**Week 4 – Lab assignment 4**
Use of R as GIS part I. Use of leaflet for mapping spatial data, use of base maps, basic geolocation in R, set map properties
**Week 5 – Lab assignment 5**
Use of R as GIS part II. Use of leaflet for mapping spatial data, use of base maps, basic geolocation in R, set map properties
**Week 6 – Lab assignment 6**
Introduction to landscape metrics. Chapter 4 in Gergel, S. E., & Turner, M. G. (2002). Learning landscape ecology: A practical guide
**Week 7 – Lab assignment 7**
Variograms and semivariograms. Chapter 5 in Gergel, S. E., & Turner, M. G. (2002). Learning landscape ecology: A practical guide
**Week 8 – Lab assignment 8**
Characterizing categorical landscape patterns. Chapter 6 in Gergel, S. E., & Turner, M. G. (2002). Learning landscape ecology: A practical guide
**Week 11 – Lab assignment 9**
Introduction to Markov Models. Chapter 8 in Gergel, S. E., & Turner, M. G. (2002). Learning landscape ecology: A practical guide
**Week 12 – Lab assignment 10**
Disturbances and network applications in landscape ecology. Chapter 11 & 12 in Gergel, S. E., & Turner, M. G. (2002). Learning landscape ecology: A practical guide

2. Learning Landscape Ecology: A Practical Guide to Concepts and Techniques (Gergel and Turner, 2001)
3. Geocomputation with R (Robin Lovelace and Muenchow, 2019)

# Chapter 1

## Introduction to Landscape Ecology

### 1.1 Requirements

**Chapter 1** - Introduction to Landscape Ecology. (Gergel and Turner, 2001)

### 1.2 Learning objectives

- Install R and R Studio in your computer
- Understand the basic datatypes and data structures in R
- Declare basic datatypes and data structures
- Do basic element-wise and vector-wise operations

### 1.3 Lab Theory

#### 1.3.1 What is R?

- R is a multiplatform programming language (*“It means that we ask the computer to do jobs.”*)
- R works in Windows, MacOS, Linux.
- R is **free**.
- Easy to learn and very powerful.
- R can do many jobs:
  - Used as a calculator.
  - R can be used to create figures such as scatter plots, bar plots, pie plots, etc.
  - R can be used to summarize data (numerically and graphically).
  - R can be used to do a lot of specific analyses.

#### 1.3.2 What is R Studio?

- RStudio is an integrated development environment (IDE) for R.
- It includes a console, syntax-highlighting editor that supports direct code execution.
- Tools for plotting, history, debugging and workspace management.
- Open source.

### 1.3.3 Installation of R and R Studio

- In this course, we run R on Windows for demonstration purposes.
- The installation of R on Windows is easy:
  - Download R (for Windows) at <https://cran.r-project.org>.
  - Run the downloaded .exe file.
  - Current version is 3.5.2 (upgraded multiple times every year).
- Installation of RStudio is also straightforward
  - Download can be found <https://www.rstudio.com/products/rstudio/download/#download>
  - Current version 1.1.463 (for windows).

## 1.4 Lab: Getting familiar with R

Installation

Calculation and variables

Create and Work With Vectors

Character and Boolean Vectors

Vector Arithmetic

## 1.5 Lab: Code practice

This section is intended to complement the concepts explained in videos. You will find all the commands used but also the expected results. Please type these codes in your RStudio Editor and start to execute (CTRL+ENTER) line by line.

### 1.5.1 Task 1: Basic variables:

```
numberStudents <- 41           # Declaration of an integer value
myFirtsName <- "Abraham"      # Declaration of a character value
myWeight <- 160.5             # Declaration of a numeric value
isUCStudent <- TRUE           # Declaration of a logical value
height <- c(152, 171.5, 165)  # Declaration of a numeric vector
mat <- matrix(
  c(14, 16, 18, 15, 22, 11),
  nrow = 3,
  ncol = 2,
  byrow = TRUE
)
# Declaration of a n-dimensional matrix
rubikCube <- array(c('green','yellow','red','blue', 'white','orange'),dim = c(3,3,6))
#Declaration of a data frame
friends <- data.frame( names = c("Thomas","Andrew","Lindsay"), gender = c("Male", "Male","Female"), hei
# See the content of your variables
print(numberStudents)         # Check the content of numberStudents
```



```
## [1] 41
```

```
print(myFirtsName)           # Check the content of myFirstName
```

```
## [1] "Abraham"
```

```
print(myWeight)              # Check the content of myWeight
```

```
## [1] 160.5
```

```
isUCStudent                  # Additional way to see the content within
```

```
## [1] TRUE
```

```
dim(mat)                     # Check the dimensions of the matrix
```

```
## [1] 3 2
```

```
print(mat)                   # Check the content of the mat matrix
```

```
##      [,1] [,2]
## [1,]   14   16
## [2,]   18   15
## [3,]   22   11
```

```
dim(rubikCube)               # Check the dimensions of the matrix
```

```
## [1] 3 3 6
```

```
print(rubikCube)             # Check your rubik content
```

```
## , , 1
##
##      [,1]      [,2]      [,3]
## [1,] "green"  "blue"    "green"
## [2,] "yellow" "white"   "yellow"
## [3,] "red"    "orange"  "red"
##
## , , 2
##
##      [,1]      [,2]      [,3]
## [1,] "blue"    "green"   "blue"
## [2,] "white"   "yellow"  "white"
## [3,] "orange"  "red"     "orange"
##
## , , 3
##
##      [,1]      [,2]      [,3]
```

```
## [1,] "green" "blue" "green"
## [2,] "yellow" "white" "yellow"
## [3,] "red" "orange" "red"
##
## , , 4
##
##      [,1]      [,2]      [,3]
## [1,] "blue" "green" "blue"
## [2,] "white" "yellow" "white"
## [3,] "orange" "red" "orange"
##
## , , 5
##
##      [,1]      [,2]      [,3]
## [1,] "green" "blue" "green"
## [2,] "yellow" "white" "yellow"
## [3,] "red" "orange" "red"
##
## , , 6
##
##      [,1]      [,2]      [,3]
## [1,] "blue" "green" "blue"
## [2,] "white" "yellow" "white"
## [3,] "orange" "red" "orange"
```

```
print(friends)           # Check list content
```

```
##      names gender height weight Age
## 1  Thomas   Male  152.0     81  42
## 2  Andrew   Male  171.5     93  38
## 3 Lindsay  Female  165.0     78  26
```

### 1.5.2 Task 2: Operations

```
a <- 1.5           # Declaration of a numeric value
b <- 4.0           # Declaration of another numeric value

a + b              # Summation
```

```
## [1] 5.5
```

```
b - a              # Substraction
```

```
## [1] 2.5
```

```
b * a              # Multiplication
```

```
## [1] 6
```

```
b / a                                # Division

## [1] 2.666667

b ** 2                               # Exponent

## [1] 16

b %% 2                               # Modulus operation

## [1] 0

# Declaration of a vector of numeric values
vector_values <- c(1, 4, 9, 12, 15, 23, 17, 13, 9)

mean(vector_values)                  # Computing the average of the set of values.

## [1] 11.44444

sum(vector_values)                   # summation of the set of values.

## [1] 103

length(vector_values)                # Computing the average of the set of values.

## [1] 9

vector_values + 3                     # Add or subtract a number to each value on the set of values.

## [1] 4 7 12 15 18 26 20 16 12

vector_values * 5                     # Multiply a number to each value on the set of values.

## [1] 5 20 45 60 75 115 85 65 45

vector_values %% 2                    # Apply a modulo operation over the set of values.

## [1] 1 0 1 0 1 1 1 1 1

vector_values[1]                      # Indexing vectors in R (find the ith element).

## [1] 1
```

```
vector_values[5]           # Suppose that we want to find the fifth element.
```

```
## [1] 15
```

```
vector_values[-2]         # Get the vector without the second element.
```

```
## [1] 1 9 12 15 23 17 13 9
```

```
vector_values[1:3]        # Get the first three elements of the vector.
```

```
## [1] 1 4 9
```

```
vector_values[c(1,3,5)]    # Get the elements 1st, 3rd and 5th of the vector.
```

```
## [1] 1 9 15
```

```
vector_values == 9         # Check if the vector values are equals to 9
```

```
## [1] FALSE FALSE TRUE FALSE FALSE FALSE FALSE FALSE TRUE
```

```
which(vector_values == 9)   # where are this value
```

```
## [1] 3 9
```

```
sum(vector_values == 9)     # How many 9's are in the vector
```

```
## [1] 2
```

### 1.5.3 Task 3. Creating vectors

```
new_vector <- 1:10          # Create a vector from 1 to 10
new_vector
```

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

```
new_vector <- rep(5, 10)     # Create a vector of 10 5's (ten times five)
new_vector
```

```
## [1] 5 5 5 5 5 5 5 5 5 5
```

```
new_vector <- seq(1, 10, 2)  # Create a vector from 1 to less equals to 10 but with step of 2
new_vector
```

```
## [1] 1 3 5 7 9
```

```
new_vector <- seq(2, 10, 2)  # Same as before but with even numbers
new_vector
```

```
## [1]  2  4  6  8 10
```

#### 1.5.4 Task 4. Operations between two or more vectors

```
vector1 <- seq(2,10,2)
vector1
```

```
## [1]  2  4  6  8 10
```

```
vector2 <- seq(11,20,2)  # Declare two vectors
vector2
```

```
## [1] 11 13 15 17 19
```

```
vector_result <- c(vector1, vector2) # Create a vector with the combination of vector1 and vector2
vector_result
```

```
## [1]  2  4  6  8 10 11 13 15 17 19
```

```
vector1 + vector2  # Addition of two vectors
```

```
## [1] 13 17 21 25 29
```

```
vector1 - vector2  # Subtraction of two vectors
```

```
## [1] -9 -9 -9 -9 -9
```

```
vector1 * vector2  # Multiplication of two vectors
```

```
## [1]  22  52  90 136 190
```

```
vector1 / vector2  # division of two vectors
```

```
## [1] 0.1818182 0.3076923 0.4000000 0.4705882 0.5263158
```

```
max(vector_result)  # vector max
```

```
## [1] 19
```

```
min(vector_result)  # vector min
```

```
## [1] 2
```

```
pmax(vector1, vector2)      # element-wise max
```

```
## [1] 11 13 15 17 19
```

```
pmin(vector1, vector2)      # element-wise min
```

```
## [1]  2  4  6  8 10
```

```
head(vector_result)         # first 5 elements
```

```
## [1]  2  4  6  8 10 11
```

## 1.6 Project 1.

- Create a vector  $x$  with 100 elements, the first 50 are 1, 2, 3, ..., 99 and the rest 50 elements are 2, 4, ..., 100.
- Create a vector  $y$  with 100 elements, the first 50 element are all 100; the rest 50 elements are 1, 2, 3, ..., 50.
- Calculate  $x+y$  and the element-wise max for  $x$  and  $y$ ; the sum of elements in  $x$  and  $y$ ; the maximum value of  $x$ .

## 1.7 Project 2.

Create a variable

```
x <- c(2, 9, 3, 7, 4, 9, 6, 6, 6, 6)
```

Compute the following:

- $(x_1 + x_2, \dots, x_{10})/10$ .
- Find  $\text{Log}_{10}(x_i)$  for each  $i$ .
- Find  $(x_i - 1.5)/2.6$  for each  $i$ .
- Find the difference between the largest and smallest values of  $x$ . (This is the range. You can use *max* and *min*.)

## 1.8 Project 3.

Suppose you keep track of one variable  $x$  at some soecific poiny in the time. The last 8 values were:

```
55311 35624 15908 36219 65499 67251 65745 34447
```

- Enter this numbers in R
  - Use commands *diff* on the data. What does it give?
  - Find the maximum number of the variable, the average number and the minimum number between the sample times.

## 1.9 Project 4.

Suppose one of your bills varies from month to month. Also suppose your year has the following monthly amounts:

50 23 42 39 47 30 31 21 12 15 55 58

- Enter this data into R using a variable called bill.
- Find the amount you spent this year on your bills.
- What is the smallest amount you spent?
- What is the largest?
- How many bills was the amount greater than \$40?
- What percentage of the total bills was this?





## Chapter 2

# Dataframes, matrices and lists in R

### 2.1 Here is a review of existing methods.

title: "Landscape ecology course: Lab session No. 2" authors: "Andres Hernandez M.Sc. - Esteban Correa M.Sc." date: "January 23, 2019" output: html\_document: toc: true toc\_float: collapsed: false smooth\_scroll: false —

#### 2.1.1 Recap

- Vector operations and generation
- Logical operations:

```
">" # for "greater than"
">=" # for "greater than or equal to"
"<" # for "less than"
"<=" # for "less than or equal to"
"==" # for "equal to"
"!=" # for "not equal to"
"&" #for "and"
"|" # for "or"
```

- Create vectors:

```
new_vector <- 1:10 # Create a vector from 1 to 10
new_vector
```

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

```
new_vector <- rep(5, 10) # Create a vector of 10 5's (ten times five)
new_vector
```

```
[1] 5 5 5 5 5 5 5 5 5 5
```

```
new_vector <- seq(1, 10, 2) # Create a vector from 1 to less equals to 10 but with step of 2
```

### 2.1.2 Introduction to Dataframes

- Usually the data is stored in a tabular form, with rows for different person and columns for different variables.
- Suppose that we have the following table:

```
plot_id <- c(3, 7, 4, 4, 7, 8, 2, 3, 7, 8, 8, 1, 7, 4, 4, 6, 19, 23, 18)
species_id <- c("DM", "DM", "DM", "DM", "DM", "DO", "PF", "OX", "DM", "PF", "DM", "PF", "DM", "PF", "DM", "DM", "DM", "DM", "DM")
sex <- c('M', 'M', 'F', 'F', 'M', 'F', 'M', 'F', 'F', 'M', 'F', 'M', 'F', 'F', 'F', 'F', 'F', 'F', 'F')
hindfoot_length <- c(35, 37, 34, 35, 35, 32, 15, 21, 36, 12, 32, 16, 34, 14, 35, 37, 35, 35, 33)
weight <- c(40, 48, 29, 46, 36, 52, 8, 22, 35, 7, 22, 9, 42, 8, 41, 37, 43, 41, 40)
study <- data.frame(plot_id, species_id, sex, hindfoot_length, weight)
```

- We can add row names to the data frame
- or change column names

```
colnames(study)<-c("ID", "NAME", "SEX", "HINDFOOT LENGHT", "WEIGHT") # Assign names to the dataframe columns
```

- Now we can summarize a column,

```
mean(study[,4]) # Compute the mean from a numeric vector
```

```
[1] 29.63158
```

```
sd(study[,4]) # Compute the standard error from a vector
```

```
[1] 8.858207
```

- Notice that we select a column by the following command:

```
(study)[,4]
```

```
[1] 35 37 34 35 35 32 15 21 36 12 32 16 34 14 35 37 35 35 33
```

- Question: how to select a row?

```
study[4,]
```

```
ID NAME SEX HINDFOOT LENGHT WEIGHT 4 4 DM F 35 46
```

- We can also select a column or a row by its name

```
study[, "NAME"] # Call column by name
```

```
[1] DM DM DM DM DM DO PF OX DM PF DM PF DM PF DM DM DM DM DM Levels: DM DO OX PF
```

To select a entry, we can do

```
study[1,2]
```

```
[1] DM Levels: DM DO OX PF
```

this is because that a table is two dimensional.

- Question: how to select the first and second column and make it a new data frame?
- Question: how to select the records (rows) for all females and make it a new data frame?

### 2.1.3 Lists

- Lists are the R objects which contain elements of different types like (numbers, strings, vectors) and another list inside it. A list can also contain a matrix or a function as its elements. List is created using `list()` function.
- Let us create a list:

```
list.example <- list( a = 1:5, b = c("Cincinnati", "Columbus", "Cleveland"), c = study)
```

- What is in the list?
- To access a list element, we can use the “\$” operator:

```
list.example$a
```

```
[1] 1 2 3 4 5
```

Or the following:

```
list.example[["a"]]
```

```
[1] 1 2 3 4 5
```

```
list.example[[1]]
```

```
[1] 1 2 3 4 5
```

- What if you use single square brackets “[” instead of “[[”?

```
list.example[2] # Access the second element of the list but not the elements within
```

```
$b [1] "Cincinnati" "Columbus" "Cleveland"
```

```
typeof(list.example[2]) # Type of element with single brackets (a list with the elements of the first p
```

```
[1] "list"
```

```
typeof(list.example[[2]]) # Type of element with double brackets
```

```
[1] "character"
```

- A data frame is a list, where all elements are the column vectors. So we can access a column of a data frame using “\$” operator:

```
study$"SEX"
```

```
[1] M M F F M F M F F M F M F F F F F Levels: F M
```

- Try the following commands:

```
length(list.example) # get the length of the list
```

```
[1] 3
```

```
length(list.example[1]) # get the length of the list returned with the elements of the first position
```

```
[1] 1
```

```
length(list.example[[1]]) # get the length of the element in the first position of the list
```

```
[1] 5
```

### 2.1.4 Matrices

A matrix is an extension of vector. It has additional dimension attributes (row and columns).

- Let us create a matrix:

```
matrix.example <- matrix(1:9, nrow = 3) # Create a matrix with 9 elements and 3 rows (columns are autom  
matrix.example
```

```
 [,1] [,2] [,3]
```

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

```
matrix.example.2 <- matrix(1:8, nrow = 2) # Create a matrix with 8 elements (1 to 8) and 2 rows.  
matrix.example.2
```

```
 [,1] [,2] [,3] [,4]
```

```
[1,] 1 3 5 7 [2,] 2 4 6 8
```

- We can change a matrix to a vector (ignore the dimension information for each entry):

```
as.vector(matrix.example.2) # get a vector (Dimension 1) with the elements of matrix.example.2
```

```
[1] 1 2 3 4 5 6 7 8
```

- How to know the dimension of a matrix:

```
dim(matrix.example.2) # Dimension of matrix.example.2
```

```
[1] 2 4
```

- What if we want to create a matrix with the elements sorted by row

```
matrix.example.byrow <- matrix( 1:9, nrow = 3, byrow = TRUE) # Create a matrix with elements by row.
matrix.example.byrow
```

```
 [,1] [,2] [,3]
```

```
[1,] 1 2 3 [2,] 4 5 6 [3,] 7 8 9 * matrices also support operations
```

```
matrix.example + matrix.example.byrow # Adding two matrices
```

```
 [,1] [,2] [,3]
```

```
[1,] 2 6 10 [2,] 6 10 14 [3,] 10 14 18
```

```
matrix.example - matrix.example.byrow # Subtracting two matrices
```

```
 [,1] [,2] [,3]
```

```
[1,] 0 2 4 [2,] -2 0 2 [3,] -4 -2 0
```

- The same logic to access the elements

```
matrix.example[1,] # Accessing row 1
```

```
[1] 1 4 7
```

```
matrix.example[,2] # Accessing row 2
```

```
[1] 4 5 6
```

```
matrix.example[1,2] # Accessing row 1,2
```

```
[1] 4
```



# Bibliography

- Gergel, S. and Turner, M. (2001). *Learning Landscape Ecology: A Practical Guide to Concepts and Techniques*, volume 83.
- Robin Lovelace, Robin, J. N. and Muenchow, J. (2019). *Geocomputation with R*, volume 1.
- Turner, M. and Gardner, R. (2015). *Landscape ecology in theory and practice: Pattern and process, second edition*.