# Introduction to R "How do you turn this thing on?"

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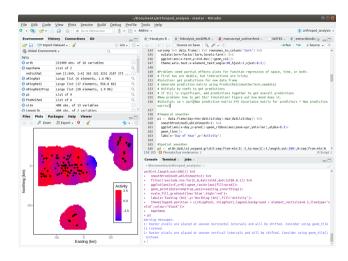
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- ► This is not a course in R programming (see here, here, here , but you do need some basic techniques

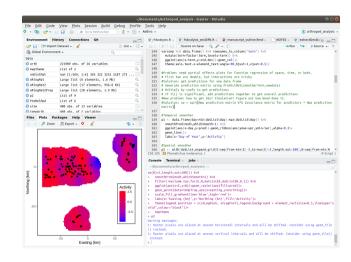
#### RStudio GUI

The Console is the main input into R (where you tell it to do things)



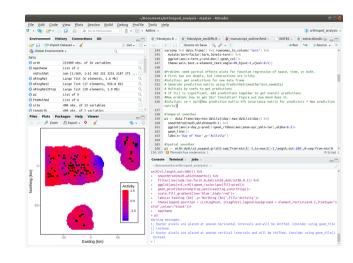
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- If you're using RStudio, 2 of the 4 panes will be dedicated to the console and scripts



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    - ▶ these are polymorphic functions: they do different things to different types of objects
  - Control flow if and for

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## **Objects**

Let's make some objects. These are all single objects:

```
myString <- "Hello world" #A string object
myNumeric <- 12345 #A numeric object
myLogical <- TRUE #A logical object</pre>
```

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▶ These are objects joined into a *vector*, joined by the function c (concatenate):

```
myCharVec <- c("I like pie", "I like cake", "I like anything you bake")
myNumVec <- c(1, 2, 3, 4, 5)
myLogVec <- c(TRUE, TRUE, FALSE, TRUE, FALSE, FALSE)</pre>
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```

► How long are each of these vectors?

```
howLong <- c(length(myCharVec), length(myNumVec), length(myLogVec))
howLong #This executes the `print` command on `howLong`
```

```
## [1] 3 5 6
```

```
myCharVec #Here's what's inside the whole thing
```

```
## [1] "I like pie" "I like cake" ## [3] "I like anything you bake"
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Single number:

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Vector of numbers

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```

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```

Results:

```
myCharVec #Here's what's inside the whole thing
```

```
## [1] "Parakeets" "Cats" "Dogs"
```

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► Logical to numeric

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Numeric to character

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Numeric to character

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```

▶ Characters to **factors**: these represent *categories* or experimental levels

```
## [1] Parakeets Cats Dogs
## Levels: Cats Dogs Parakeets
```

#### **Dataframes**

## 2

## 3

Cats

Dogs

TRUE

FALSE

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```
summary(myDF) #This function summarizes each column
```

```
##
    stringCol
                        numCol
                                   logCol
   Length:3
             Min. :1.0 Mode :logical
##
   Class:character 1st Qu.:1.5 FALSE:1
                    Median: 2.0 TRUE: 2
##
   Mode :character
                    Mean :2.0
##
                    3rd Qu.:2.5
##
##
                    Max · 3 0
```

```
myDF[1, 2]
## [1] 1
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myDF$numCol #This gets all of the column 'numCol'
## [1] 1 2 3
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Like other objects, you can alter dataframes

► You can add columns

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You can also alter columns in place, or elements within columns

```
myDF$numCol <- (myDF$numCol)^2 #Square of numCol
myDF$numCol[3] <- myDF$numCol[3] - myDF$numCol[2] #Subtracts 2nd from 3rd</pre>
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We'll learn easier ways of doing this next week using the tidyverse

### Reading csv files

▶ One common practice is to read in your own dataframe from a csv file <sup>2</sup>

```
testDat <- read.csv("test_results.csv") #Path to csv file
head(testDat) #head shows only the first 6 rows of dataframe
```

```
##
     Concentration Treatment Lab.Member Time.of.Day
## 1
               2.9
                     Control
                                   Will
                                            Morning
## 2
               3.2
                     Control
                                   Will
                                            Morning
               3.6
                     Control
## 3
                                   Will
                                            Morning
               5.6
## 4
                                   Will
                                            Morning
               6.8
## 5
                                   Will
                                            Morning
## 6
               7.0
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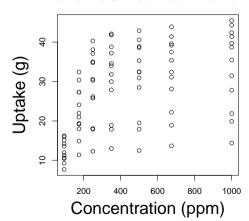
- R can't handle spaces or other special characters in the column headers (replaces them with periods). It also tries to guess the proper data type for each column
- ► Make sure that the csv file is in your working directory

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## **Plotting**

The plot command is useful for quickly looking at sets of data. The following CO2 dataset is built-in to R.3

### **Plot of CO2 Concentrations**



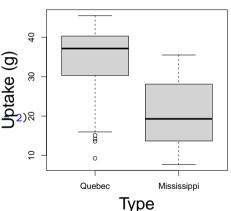
<sup>&</sup>lt;sup>3</sup>To see others, type data() in the console

### More Plotting

The boxplot command can summarize continuous and categorical data

```
# Boxplot uses a formula rather than
# x,y vectors

# Formula: uptake depends on (~) Type
boxplot(CO2$uptake ~ CO2$Type, xlab = "Type",
ylab = "Uptake (g)", cex.lab = 2, cex.main 22)%
```



## First challenge

Your supervisor has just given you a dataset (test\_results.csv) recorded by two undergrads. However, these undergrads were in a hurry and have made some mistakes:

- Make a script in R, and use this to record what you do
- ► Read the *csv* file and fix any mistakes. Bonus if you do this without using Excel!
- ▶ Plot the concentration data by treatment group, then plot it for each undergrad. Does there look like much of a difference?
- ► Some useful commands: read.csv, boxplot, is.na, as.factor, summary

#### **Functions**

► Functions take objects as **arguments** (input) and return other **objects** (output)

```
myNumVec <- c(1, 2, 3, 4, 5)

meanVec <- mean(myNumVec) #Arithmetic mean (average)

sdVec <- sd(myNumVec) #Standard deviation (sqrt(variance))

meanSdVec <- c(meanVec, sdVec) #Joins mean and SD into a vector

meanSdVec
```

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▶ If you can't remember how a command works, use ? in the console to access the help files (e.g. ?median)

#### Homemade Functions

➤ You can make your own functions! This is useful if you have to do the same thing to many different input objects.

```
myFun <- function(input) {
    # Takes a vector of numbers
    A <- mean(input) #Take the mean of INPUT
    B <- sd(input) #Take the SD of INPUT
    C <- c(A, B) #Join A and B into a vector C
    return(C) #Return (output) C, then end the function
}
myFun(myNumVec) #Same as previous slide</pre>
```

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```

## [1] 3.000000 1.581139

► The objects inside of functions (A, B, C in the one above) disappear after the function runs. However, functions can see objects in the outer environment, so beware of the Steve Problem\*

### Summary statistics

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Using the CO2 plant example, how does uptake differ between Type?

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# Split up uptake by Type and Treatment, then take the mean
tapply(CO2$uptake, list(CO2$Type, CO2$Treatment), mean)
```

```
## nonchilled chilled
## Quebec 35.33333 31.75238
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```

► Typing "CO2" over and over again is annoying. You can use with to avoid repeating the name of the dataframe (avoid using attach)

```
# Runs command inside the name space of the CO2 object
with(CO2, tapply(uptake, list(Type, Treatment), sd))
```

#### if statements

▶ R can be told to do things only *if* certain conditions apply. This is useful inside of functions for error handling:

```
myFun2 <- function(x) {</pre>
    xClass <- class(x) #What class is x? (Numeric, character, boolean)
    if (xClass == "character") {
        # == means 'are these things equal'?
        return("This is a string") #If x is a character, returns a message
    } else {
        return(mean(x)) #If x isn't a character, returns the mean of x
myFun2(myCharVec)
```

```
## [1] "This is a string"
myFun2(myNumVec)
```

## [1] :

### for loops

▶ R can be told to do things repeatedly, using an *index* inside a loop:

```
classVec <- rep(0, 10) #Storage vector of zeros, 10 long
classVec[c(1, 2)] <- 1 #Set first two slots to 1

# Each time the loop repeats, i will take on values 3 to 10

for (i in 3:length(classVec)) {
    # ith slot of classVec becomes the sum of the previous two slots
    classVec[i] <- classVec[i - 1] + classVec[i - 2]
}</pre>
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}</pre>
```

classVec #First 10 numbers in the Fibbonaci sequence

```
## [1] 1 1 2 3 5 8 13 21 34 55
```

Population growth models are common in ecology, and usually often take the form  $n_t = n_{t-1} + rn_{t-1}$ , where n is the number of critters at some time point t, and r is the change in n from one point to the next (r = 0: no change).

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  - Exponential growth:  $n_t = n_{t-1}(1+r)$
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- Make a simple plot of your results

#### Lists

- Lists look similar to vectors, but can hold anything in each slot, including other lists.
- ► LOTS of things in R (e.g. model output) are specially-structured lists at their core

```
myList <- list(charSlot = myCharVec,
    numSlot = myNumVec, logSlot = myLogVec,
    dfSlot = myDF)</pre>
```

```
## $charSlot
  [1] "Parakeets" "Cats"
                                "Dogs"
##
## $numSlot
## [1] 1 2 3 4 5
##
## $logSlot
## [1]
        TRUE
              TRUE FALSE
                          TRUE FALSE FALSE
##
## $dfSlot
##
     stringCol numCol logCol
## 1 Parakeets
                        TRUE
## 2
          Cats
                        TRUE
## 3
          Dogs
                       FALSE
```

### **Accessing Lists**

Lists can be accessed numerically or logically, or by their name slots:

```
myList[[2]] #Needs 2 square brackets to isolate object
## [1] 1 2 3 4 5
myList[["numSlot"]]
## [1] 1 2 3 4 5
myList$numSlot
## [1] 1 2 3 4 5
myList[[4]][, 3] #Same as myList$dfSlot$logCol
## [1]
       TRUE TRUE FALSE
```

# lapply and sapply

Loops can be slow: using a **functional** is generally better (if each step doesn't depend on the previous one)

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```
input <- c(1, 2, 3) #Input vector
pFun <- function(N, type = "normal") {
    if (type == "poisson")
        rpois(N, 3) else rnorm(N, 3)
}
lapply(input, pFun, type = "poisson") #Generates a list of N Poisson random numbers</pre>
```

```
##
## [[2]]
## [1] 5 1
##
## [[3]]
## [1] 1 7 4
```

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

### lapply and sapply

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```
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lapply(input, pFun, type = "poisson") #Generates a list of N Poisson random numbers</pre>
```

```
## [[1]]

## [1] 5

##

## [[2]]

## [1] 5 1

##

## [[3]]

## [1] 1 7 4
```

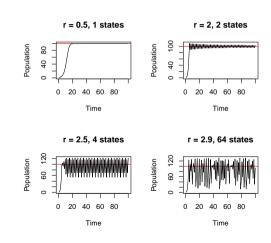
Different output formats: lapply uses a list, sapply uses a vector (or matrix)

## Third challenge

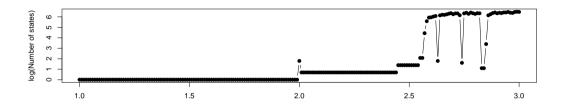
The discrete logistic growth equation has some interesting mathematical properties (see here and here). At certain values of r the population converges to the carrying capacity K, while at other values it cycles between multiple values.

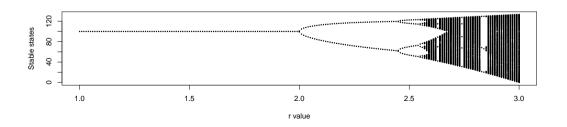
Using *lapply*, *sapply* and the logistic function you wrote for the second challenge:

- Write a script that iterates across a range of r values: 1 < r < 3
- At each *r* value, how many unique values exist?
- ▶ Keep K at 100 and  $n_0$  at 1 for now. Use a simulation time of t = 1000, but discard the first 100 points



# Complex behaviour from a simple system





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- Take a look at a data repository like Dryad