

# Introduction to R

Project Set-Up

Data Management

Coding

Alexander Hurley - November 27, 2015

# Introduction

# Why & How?

- Introduction
  - Project Set-Up
  - Data Management
  - Coding
  - Conclusion
-

# Why & How?

- Introduction
  - Project Set-Up
  - Data Management
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  - Conclusion
- 

provide knowledge and tools for developing an individual  
**project management framework** that allows **reproducing** your research  
and **sharing** it with others

# Why & How?

- Introduction
  - Project Set-Up
  - Data Management
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  - Conclusion
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# Why & How?

- Introduction
  - Project Set-Up
  - Data Management
  - Coding
  - Conclusion
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- RStudio
- PackRat
- Git and GitHub
- Online Resources

# Project Set-Up



# Why invest time?

- Introduction
  - Project Set-Up
  - Data Management
  - Coding
  - Conclusion
-

# Why invest time?

● Introduction   ● Project Set-Up   ● Data Management   ● Coding   ● Conclusion

---

- Develop and manage your workflow
- Save time and energy in the long run
- Structure allows 'outsiders' to understand quicker

# Why invest time?

● Introduction   ● Project Set-Up   ● Data Management   ● Coding   ● Conclusion

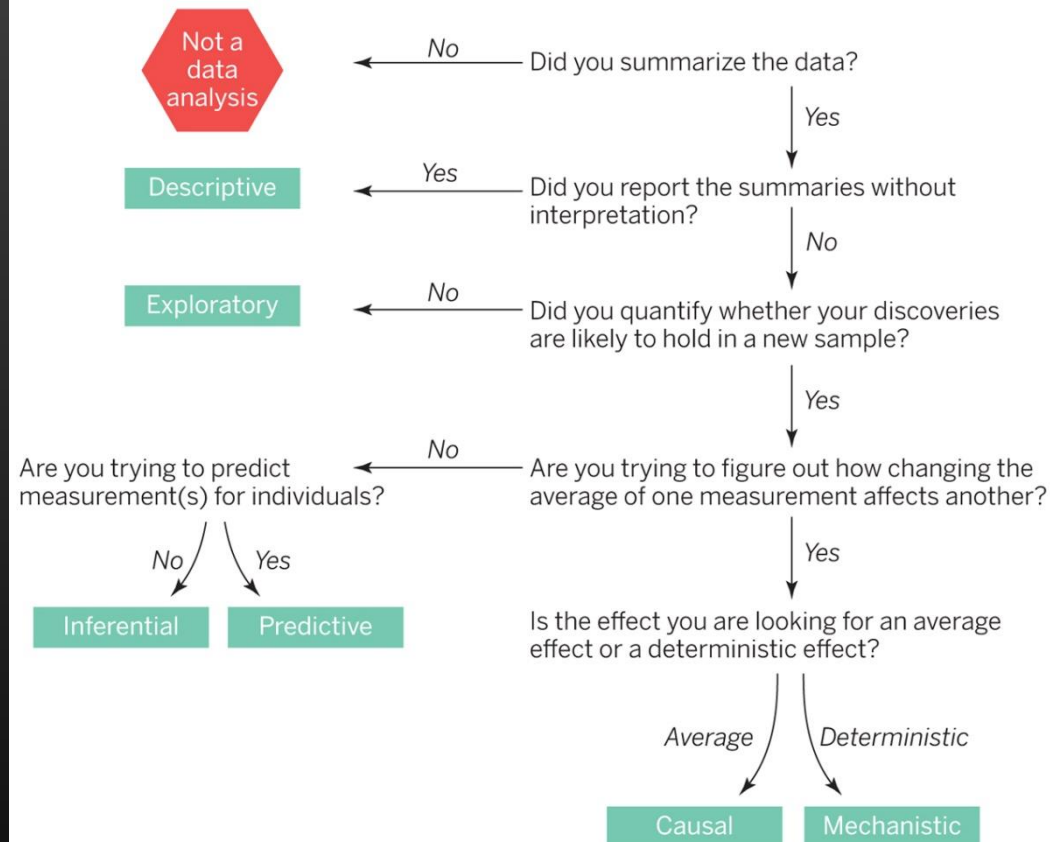
---

- Develop and manage your workflow
- Save time and energy in the long run
- Structure allows 'outsiders' to understand quicker
- Supports and reinforces design of data analyses and interpretation

# Why invest time?

- Introduction
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## Data analysis flowchart



Jeffery T. Leek, and Roger D. Peng. *Science* 2015;347:1314-1315

# Components

- Introduction
  - Project Set-Up
  - Data Management
  - Coding
  - Conclusion
-

# Components

● Introduction   ● Project Set-Up   ● Data Management   ● Coding   ● Conclusion

---

- Raw data
- Tools for data processing (i.e. cleaning)
- Tools for analyses

# Components

● Introduction   ● Project Set-Up   ● Data Management   ● Coding   ● Conclusion

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- Raw data
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Storage (tools, scripts, outputs)

# Components

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- Raw data
- Tools for data processing (i.e. cleaning)
- Tools for analyses



Storage (tools, scripts, outputs)

- Sharing / publishing results



# RStudio helps..

- Introduction
  - Project Set-Up
  - Data Management
  - Coding
  - Conclusion
- 

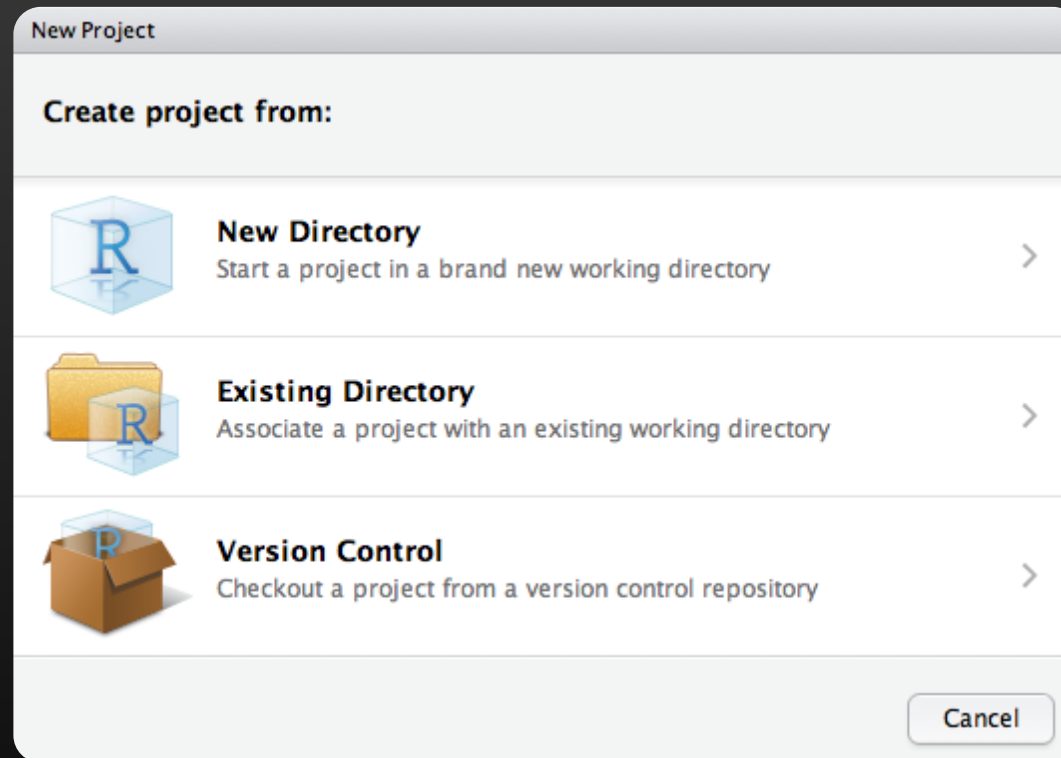
Use built-in project functionality:

# RStudio helps..

● Introduction   ● Project Set-Up   ● Data Management   ● Coding   ● Conclusion

Use built-in project functionality:

File >

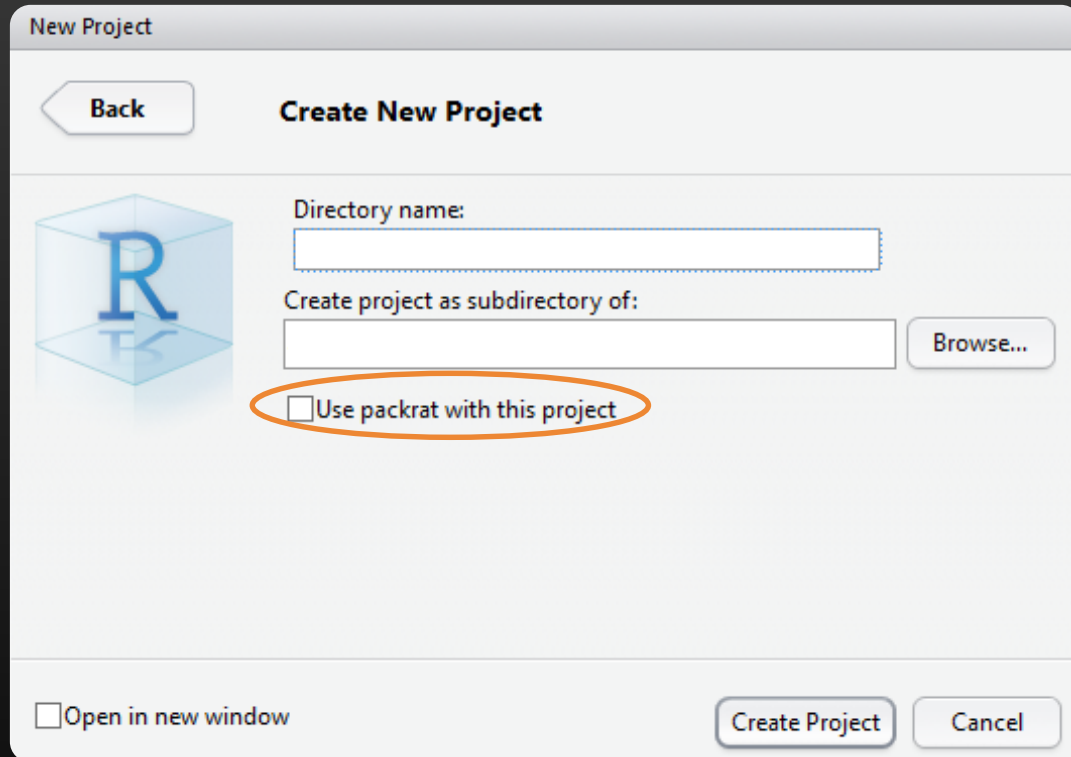


# RStudio helps..

● Introduction   ● Project Set-Up   ● Data Management   ● Coding   ● Conclusion

Use built-in project functionality:

File >



The image shows the 'New Project' dialog box in RStudio. The title bar reads 'New Project'. Inside the dialog, there is a 'Back' button on the left. The main heading is 'Create New Project'. On the left side, there is a large blue 'R' logo. To the right of the logo, there are two text input fields. The first is labeled 'Directory name:' and the second is labeled 'Create project as subdirectory of:'. To the right of the second input field is a 'Browse...' button. Below these fields, there is a checkbox labeled 'Use packrat with this project', which is circled in orange. At the bottom left, there is a checkbox labeled 'Open in new window'. At the bottom right, there are two buttons: 'Create Project' and 'Cancel'.

# RStudio helps..

● Introduction   ● Project Set-Up   ● Data Management   ● Coding   ● Conclusion

---

Use built-in project functionality:

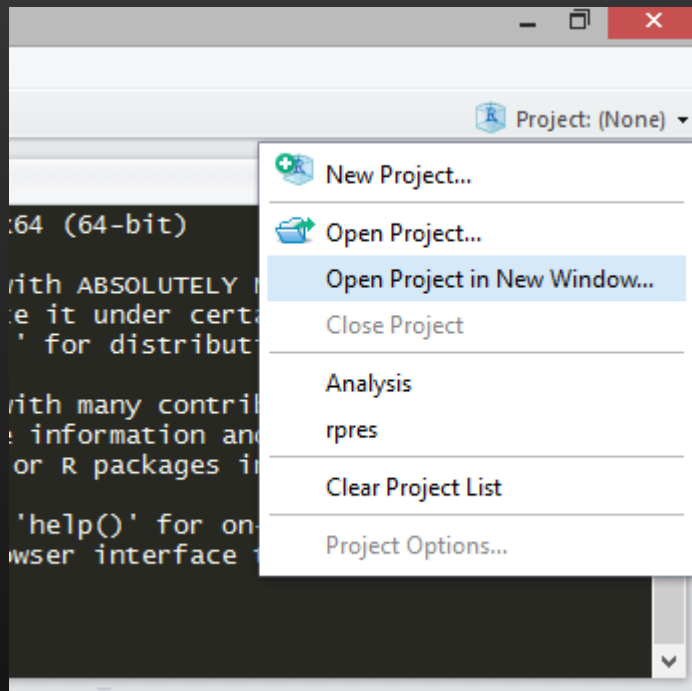
- *.Rprofile* automates tasks
- *.Rdata* keeps a snapshot of your working directory
- Sets working directory automatically
- Reloads open scripts, history, project-specific settings
- ...

See: <https://support.rstudio.com/hc/en-us/articles/200526207-Using-Projects>

# RStudio helps..

- Introduction
- Project Set-Up
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Use built-in project functionality:



Keep track of multiple projects  
**at the same time!**

# RStudio helps..

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- Conclusion

## *.RProfile:*

```
source('~\your_functions.R') # Smart working directory (relative paths)
                             # Auto-load your own functions / exec. scripts

# First function called when you start R:
.First <- function(){
  cat("\nHello UseR!\n")
}

# Last function called when you quit R:
.Last <- function(){
  cat("\nBye UseR!\n")
}
```

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**Transfer between systems!**

# PackRat helps..

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  - Data Management
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New versions of R & packages..



# PackRat helps..

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New versions of R & packages..

- Sharing your analysis with someone = nightmare?

# PackRat helps..

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- Re-running analysis after x years = nightmare?

# PackRat helps..

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New versions of R & packages..

- Sharing your analysis with someone = nightmare?
- Re-running analysis after x years = nightmare?
- Updating packages half-way through analysis = **disaster**!?

# PackRat helps..

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New versions of R & packages..

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**PackRat** acts as a safe, storing packages in the versions you used

# PackRat helps..

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New versions of R & packages..

- Sharing your analysis with someone = nightmare?
- Re-running analysis after x years = nightmare?
- Updating packages half-way through analysis = **disaster**!?

**PackRat** acts as a safe, storing packages in the versions you used



isolated, portable & reproducible.

See: <https://rstudio.github.io/packrat/>

# PackRat helps..

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## Getting and using PackRat:

(Requires Rtools; see: <https://support.rstudio.com/hc/en-us/articles/200486498-Package-Development-Prerequisites>)

```
install.packages('packrat')

## Tell packrat where to work
packrat::init ('~/projects/experiment-1') # creates the 'safe' (local library)

## Start the job
install.packages('reshape2')
library('reshape2') # load package

## Tell packrat to save the packages your working with
packrat::snapshot()

## Check what packrat's doing
packrat::status() # shows unused and missing packages from the 'safe'
```

# Git and GitHub help..

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# Git and GitHub help..

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---

- Saving your work (scripts, analysis) is important.
- But: Overwriting a file can be catastrophic.



# Git and GitHub help..

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> **Git** is a *version control system* keeping track of files in your project

# Git and GitHub help..

● Introduction   ● Project Set-Up   ● Data Management   ● Coding   ● Conclusion

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- Saving your work (scripts, analysis) is important.
- But: Overwriting a file can be catastrophic.

- **Git** is a *version control system* keeping track of files in your project
- **Git** can *mirror* your project across computers and operating systems

See: <https://help.github.com/articles/good-resources-for-learning-git-and-github/>

See: <https://try.github.io/levels/1/challenges/1>

# Git and GitHub help..

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## Git vocabulary:

Directory	=	Local file storage
Repository	=	Central file storage tracked by Git
Save	=	Local operation
Commit	=	Tell Git to take a snapshot
Push	=	Move snapshot to repository
Clone	=	Mirror repository to your local directory

# Git and GitHub help..

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**GitHub** is a user-friendly way of implementing **Git**

See: <https://github.com/>

# Git and GitHub help..

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## GitHub provides:

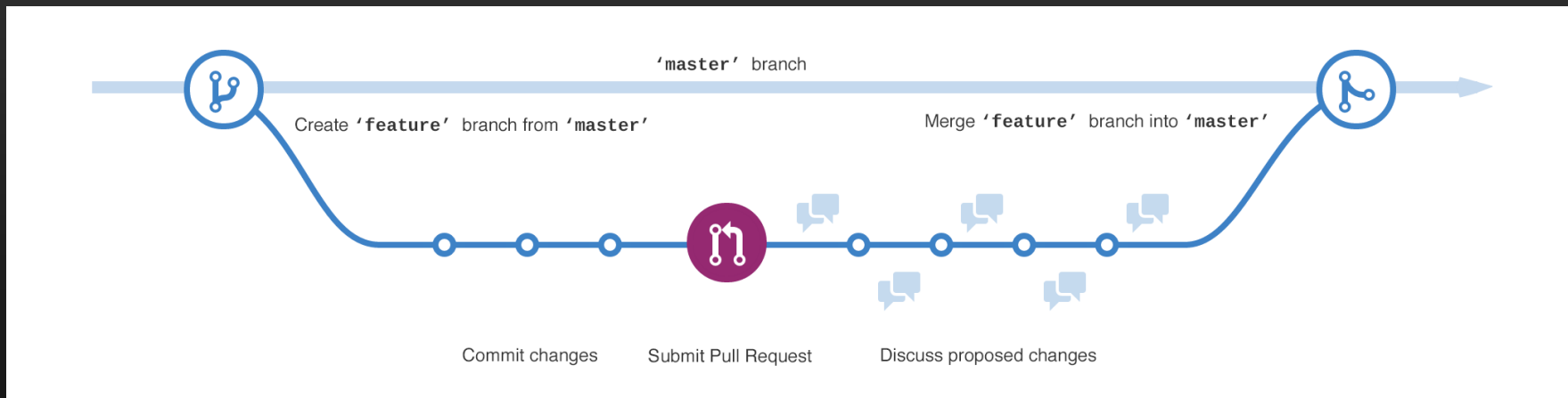
- Easy-to-use GUI
- track changes, discuss issues and collaborate

# Git and GitHub help..

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## GitHub provides:

- Easy-to-use GUI
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See: <https://guides.github.com/activities/hello-world/>

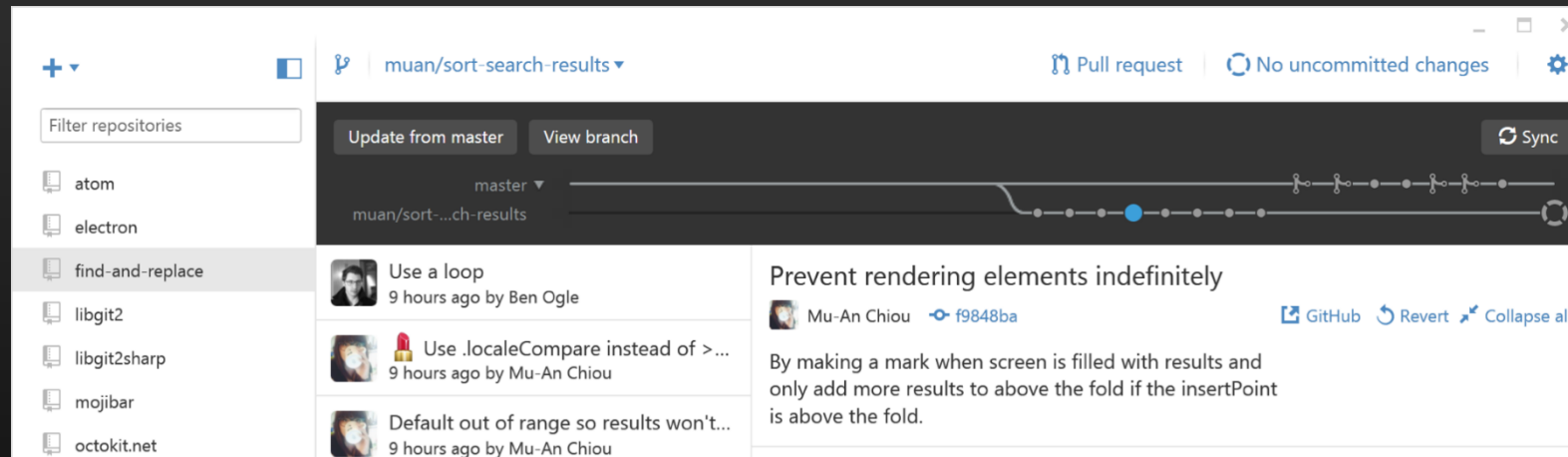
See: <https://github.com/the-Hull/Diss> (personal repo)

# Git and GitHub help..

● Introduction   ● Project Set-Up   ● Data Management   ● Coding   ● Conclusion

## GitHub provides:

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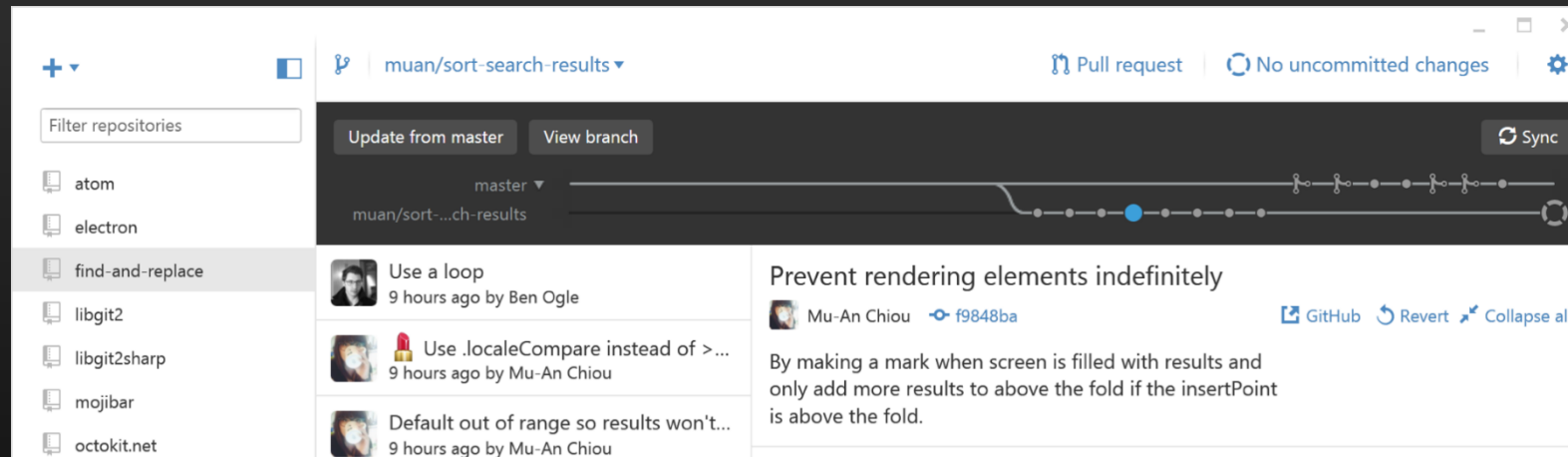
See: <https://desktop.github.com/>

# Git and GitHub help..

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- Easy-to-use GUI
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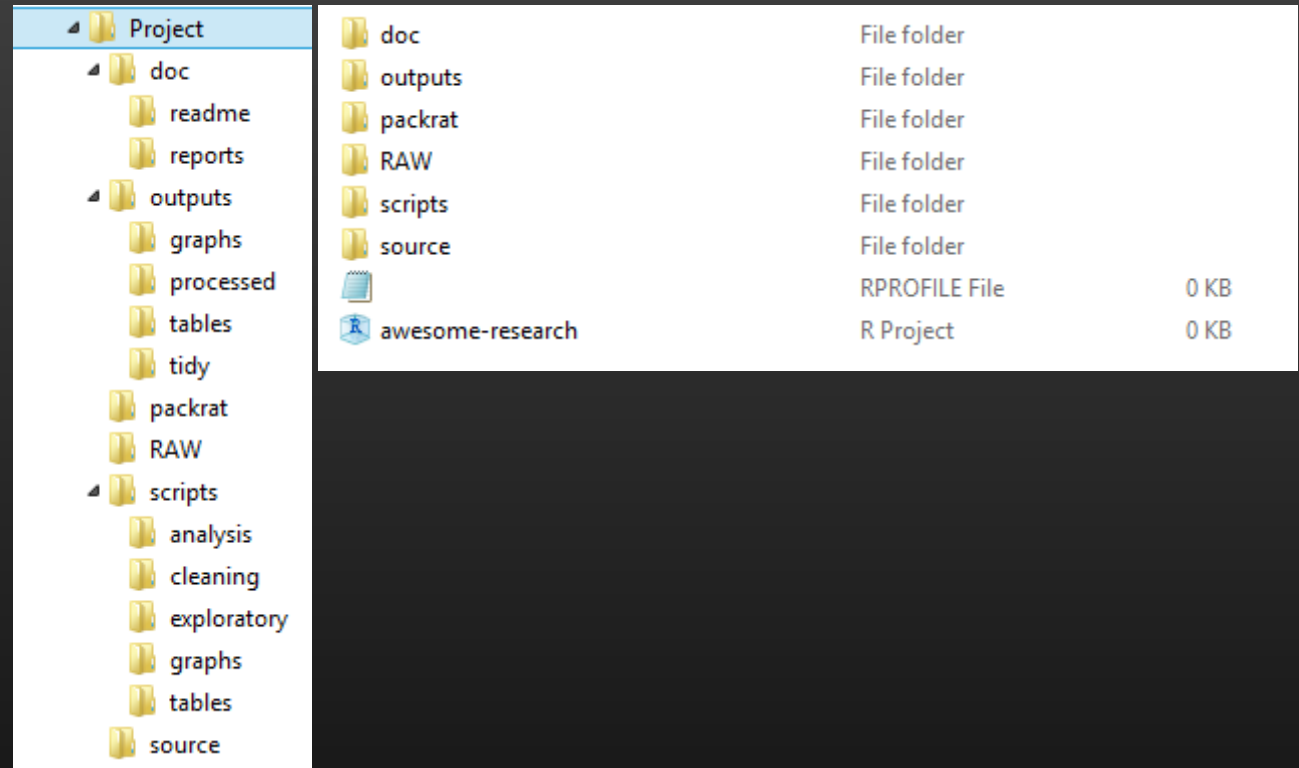


# Hierarchy & Order help..

● Introduction   ● Project Set-Up   ● Data Management   ● Coding   ● Conclusion

## Personal preference of a project:

- RAW
- Scripts
  - cleaning
  - exploratory
  - analysis
  - graphs
  - tables
- Source (helper functions)
- Outputs
  - tidy data
  - processed
  - graphs
  - tables
- Documentation (readme files, reports)
- **PackRat** folders



# Data Management

# Data.. set?

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Components:

# Data.. set?

● Introduction   ● Project Set-Up   ● Data Management   ● Coding   ● Conclusion

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## Components:

- Raw Data
- Scripts (processing)
- Tidy Data

# Data.. set?

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---

## Components:

- Raw Data
- Scripts (processing)
- Tidy Data
- Code book

# Data.. set?

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Code Book:

# Data.. set?

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---

## Code Book:

- Information on **origin** and **type** of raw data (incl. missing values, units)
- Detail choices and steps made for processing data (e.g. renaming, averaging)
- Outline study design (e.g. for identifying confounding factors)

# Data.. set?

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## Code Book:

- Information on **origin** and **type** of raw data (incl. missing values, units)
- Detail choices and steps made for processing data (e.g. renaming, averaging)
- Outline study design (e.g. for identifying confounding factors)
- Formats: *\*.docx*, *\*.txt*, *\*.Rmd* (RMarkdown)

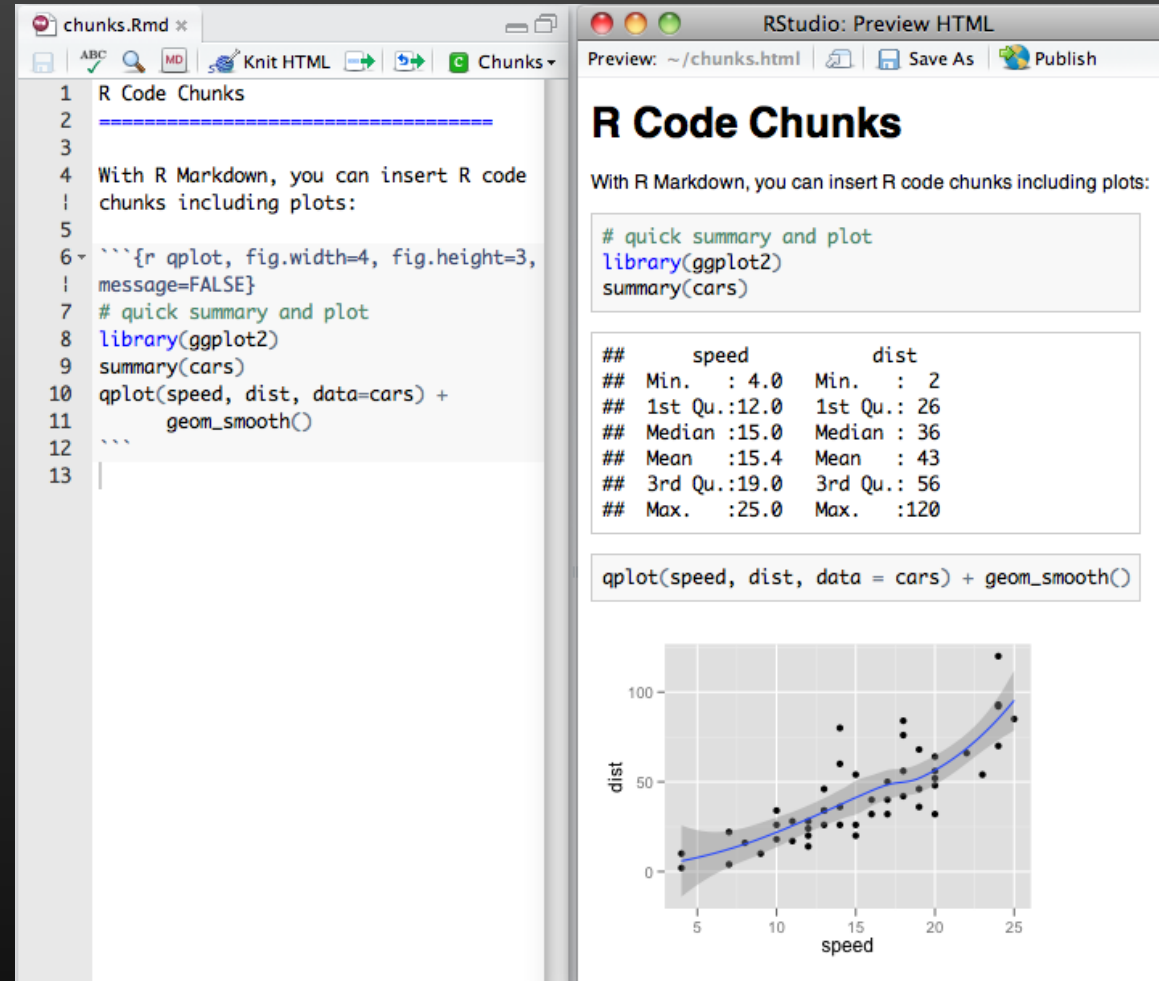
See: <http://rmarkdown.rstudio.com/>



# Data.. set?

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## Code Book:



The screenshot displays the RStudio interface with two windows. The left window, titled 'chunks.Rmd', shows R code chunks for a document. The right window, titled 'RStudio: Preview HTML', shows the rendered HTML output of the same code.

**Left Window (chunks.Rmd):**

```
1 R Code Chunks
2 =====
3
4 With R Markdown, you can insert R code
5 chunks including plots:
6
7 ```{r qplot, fig.width=4, fig.height=3,
8   message=FALSE}
9 # quick summary and plot
10 library(ggplot2)
11 summary(cars)
12 qplot(speed, dist, data=cars) +
13   geom_smooth()
```

**Right Window (RStudio: Preview HTML):**

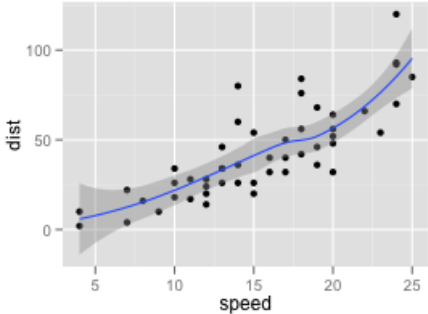
**R Code Chunks**

With R Markdown, you can insert R code chunks including plots:

```
# quick summary and plot
library(ggplot2)
summary(cars)
```

##	speed	dist
##	Min. : 4.0	Min. : 2
##	1st Qu.:12.0	1st Qu.: 26
##	Median :15.0	Median : 36
##	Mean :15.4	Mean : 43
##	3rd Qu.:19.0	3rd Qu.: 56
##	Max. :25.0	Max. :120

```
qplot(speed, dist, data = cars) + geom_smooth()
```



See: <http://rmarkdown.rstudio.com/>

# Data.. storage?

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Think ahead:

# Data.. storage?

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---

Think ahead:

- Use clear, systematic names with **common identifiers**:

# Data.. storage?

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---

Think ahead:

- Use clear, systematic names with **common identifiers**:

*exp1-plot1.csv*  
*exp1-plot2.csv*  
*exp2-plot1.csv*  
*exp2-plot2.csv*

# Data.. storage?

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Think ahead:

- Use clear, systematic names with **common identifiers**:

*exp1-plot1.csv  
exp1-plot2.csv  
exp2-plot1.csv  
exp2-plot2.csv*

```
## get a list of all files of exp1
exp1_list <- list.files(path = '~/raw',
                       pattern = "exp1")

## load all data from exp1 into the working environment
## and do the same operation...

## use file names to label objects
exp_labels <- gsub(pattern = '-plot.*', replacement = "", x = exp1_list)
```

# Data.. storage?

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# Data.. storage?

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Keeping your work organized in R:

**environments** are 'folders' in R's working environment

# Data.. storage?

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Keeping your work organized in R:

**environments** are 'folders' in R's working environment

*Working  
Environment*

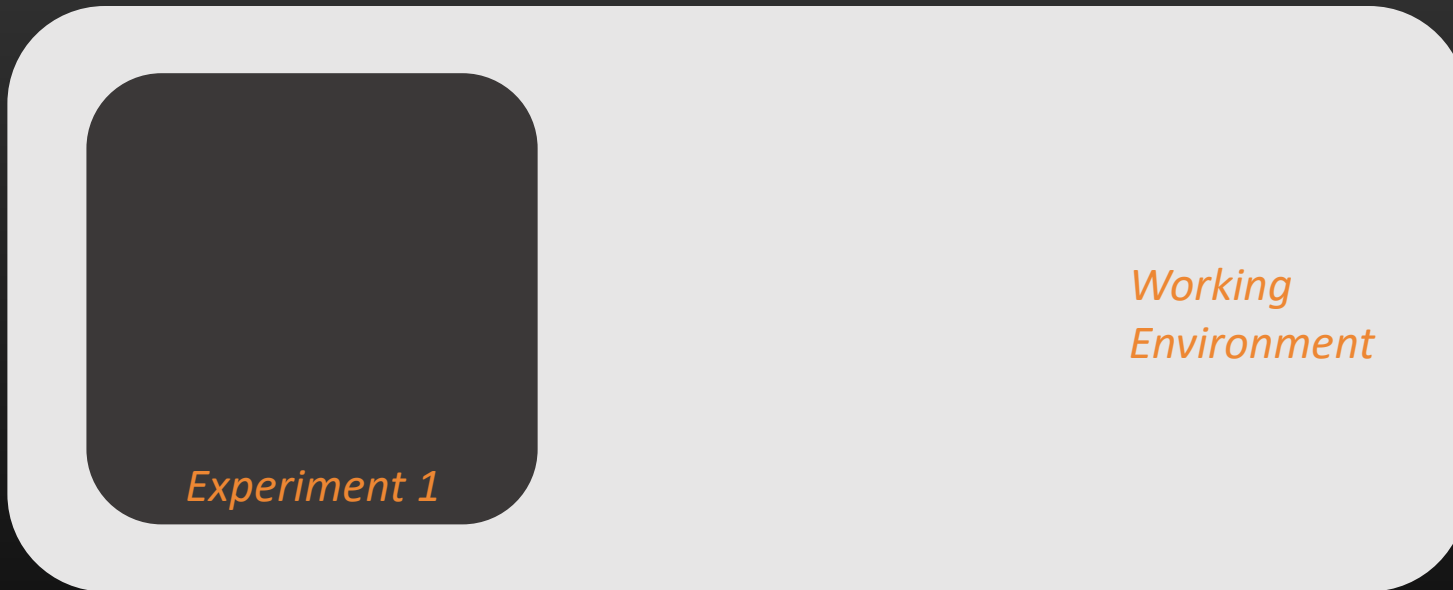


# Data.. storage?

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Keeping your work organized in R:

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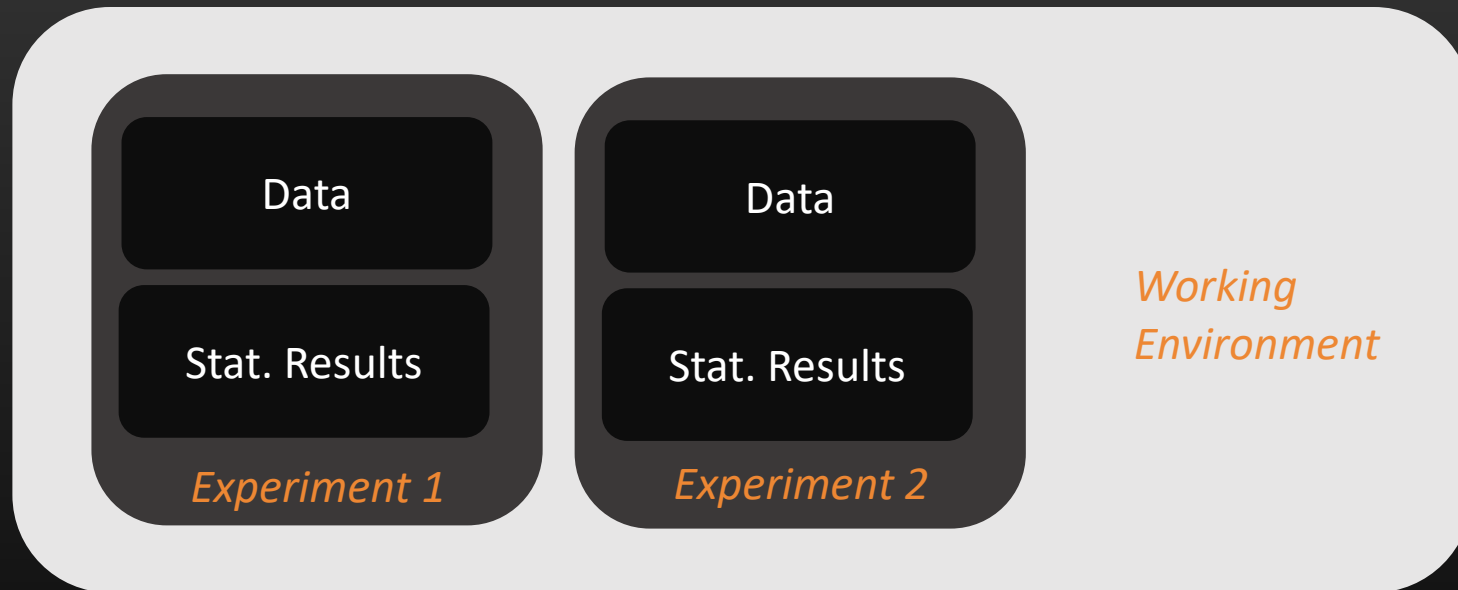


# Data.. storage?

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Keeping your work organized in R:

**environments** are 'folders' in R's working environment



Important functions:

*new.env()*

*assign()*

*get()*

*save()* / *load()*



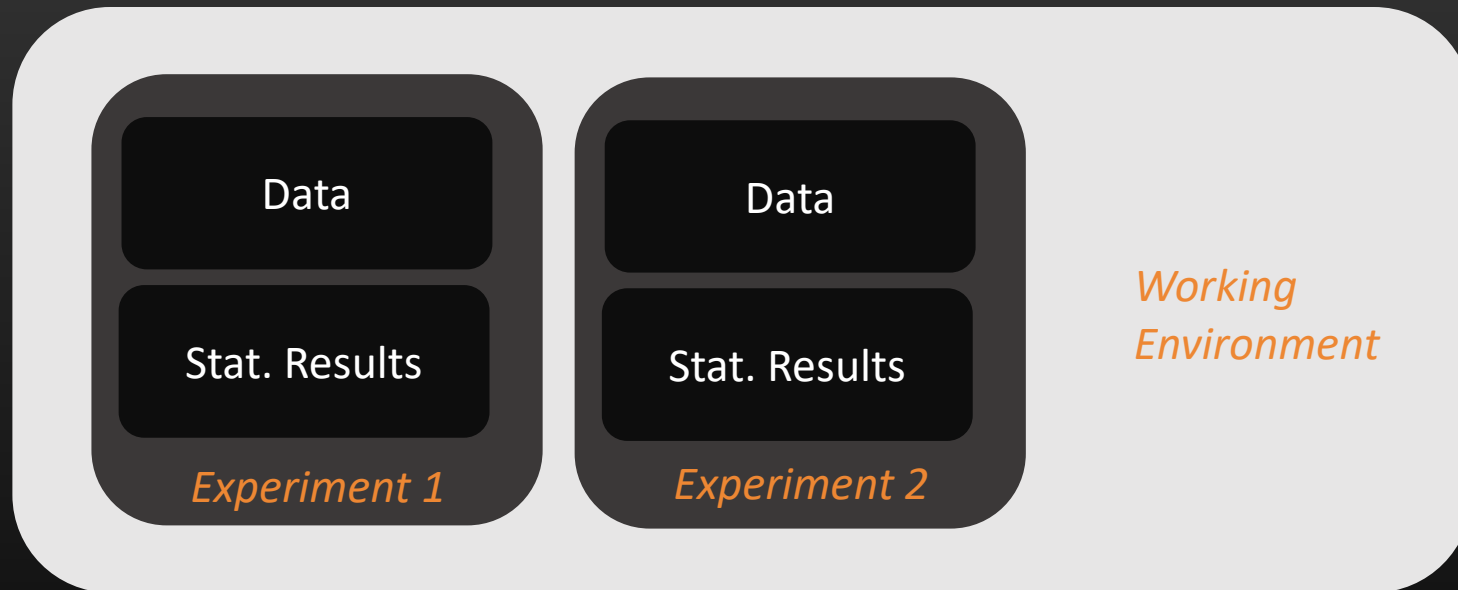
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BIRMINGHAM

# Data.. storage?

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## Keeping your work organized in R:

**environments** are ‘folders’ in R’s working environment



Important functions:

*new.env()*  
*assign()*  
*get()*  
*save()* / *load()*

Benefits:

- Clean working env.
- Less chance of confusing objects
- Repeating processes (e.g. loops)
- Saving & re-using outputs

See: <http://adv-r.had.co.nz/Environments.html>

# Data.. storage?

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## Keeping your work organized in R:

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# Data.. help?

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Useful packages and links for data processing:

- dplyr
- reshape2

See: <http://www.rstudio.com/wp-content/uploads/2015/02/data-wrangling-cheatsheet.pdf> (!)

# Coding



# Expressing yourself

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- Writing code is like using a ‘natural’ language

# Expressing yourself

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- Writing code is like using a 'natural' language
  - syntax, semantics (and beauty?)
  - learning is time consuming
  - rewarding

# Guidelines

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## Example:

```
## This script is part of awesome-research phase 1. It takes raw data gathered from exp 1
## and exp 2 (g/cm), calculates their means, and saves the result for later use in output/processed.

exp1 <- c(1, 2, 3) #loads raw data
exp2 <- c(4, 5, 6)
exp.names <- c("exp1", "exp2") # consider using function ls() to obtain names in working env.

# Necessary functions -----
averageExp <- function(experiments){
  ## function takes character vector of experiment names
  ## names and averages outcomes. Results are stored in new vector

  data.temporary <- mget(experiments,
                        envir = .GlobalEnv) # get values from environment

  result <- sapply(data.temporary, mean)
}

# Calculation -----
exp.means <- averageExp(exp.names)

# Saving Data -----
save(file = "~/output/processed/exp.means.rda", x = exp.means)
```

# Guidelines

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## Example:

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Clear description at top

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```
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Clear description at top

Useful comments  
when needed

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  ## names and averages outcomes. Results are stored in new vector

  data.temporary <- mget(experiments,
                          envir = .GlobalEnv) # get values from environment

  result <- sapply(data.temporary, mean)

# Calculation -----
exp.means <- averageExp(exp.names)

# Saving Data -----
save(file = "~/output/processed/exp.means.rda", x = exp.means)
```

Clear description at top

Useful comments  
when needed

Separated into chunks

# Guidelines

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## Example:

```
## This script is part of awesome-research phase 1. It takes raw data gathered from exp 1
## and exp 2 (g/cm), calculates their means, and saves the result for later use in output/processed.

exp1 <- c(1, 2, 3) #loads raw data
exp2 <- c(4, 5, 6)
exp.names <- c("exp1", "exp2") # consider using function ls() to obtain names in working env.

# Necessary functions -----
averageExp <- function(experiments){
  ## function takes character vector of experiment names
  ## names and averages outcomes. Results are stored in new vector

  data.temporary <- mget(experiments,
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  result <- sapply(data.temporary, mean)

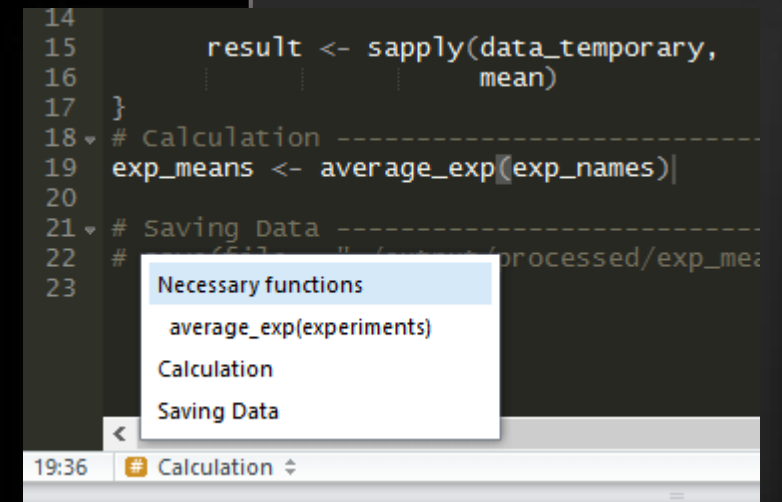
# Calculation -----
exp.means <- averageExp(exp.names)

# Saving Data -----
save(file = "~/output/processed/exp.means.rda", x = exp.means)
```

Clear description at top

Useful comments  
when needed

Separated into chunks



The screenshot shows a code editor with R code. A menu is open over the code, listing the following items:

- Necessary functions
- average\_exp(experiments)
- Calculation
- Saving Data

The code in the background includes lines 14 through 23, showing the calculation of means for two experiments and saving the results.

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## Example:

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Clear description at top

Useful comments  
when needed

Separated into chunks

Defined function



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exp.means <- averageExp(exp.names)

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Clear description at top

Useful comments  
when needed

Separated into chunks

Defined function

Useful object names

# Guidelines

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## Example:

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Clear description at top

Useful comments  
when needed

Separated into chunks

Defined function

Useful object names

Line wrapping

# Guidelines

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- 

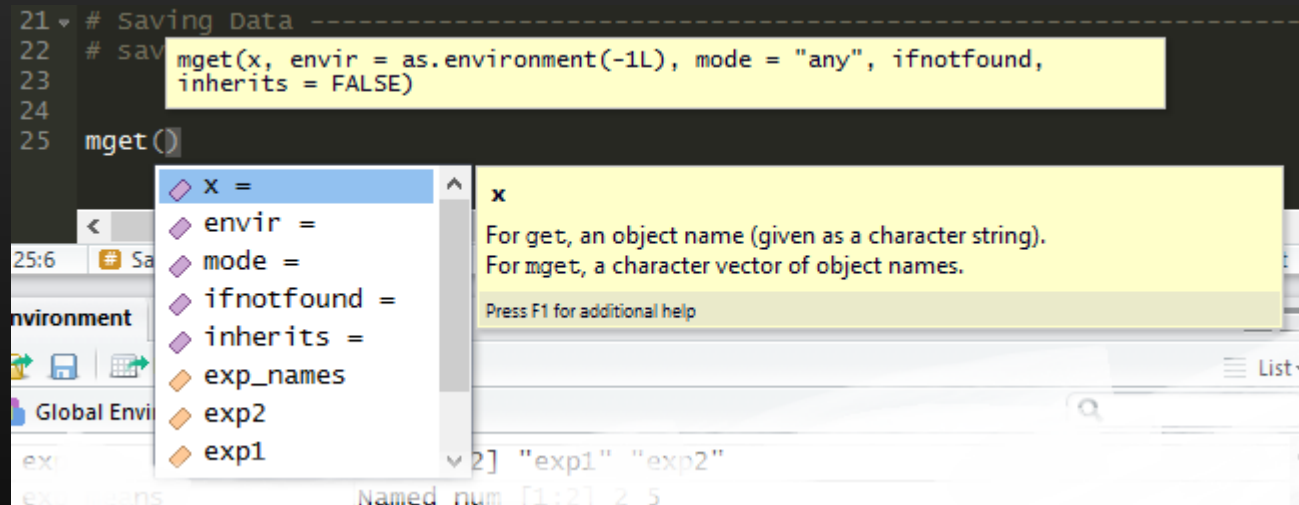
- Other Guidelines for R:
  - naming (funCtions, var.iables, CONSTANTS)
  - indention (2 to 4 spaces)
  - spacing for operators (4 + 4; x[1, ])
  - assign with <- **not** =

See: <https://google.github.io/styleguide/Rguide.xml>

# RStudio helps again..

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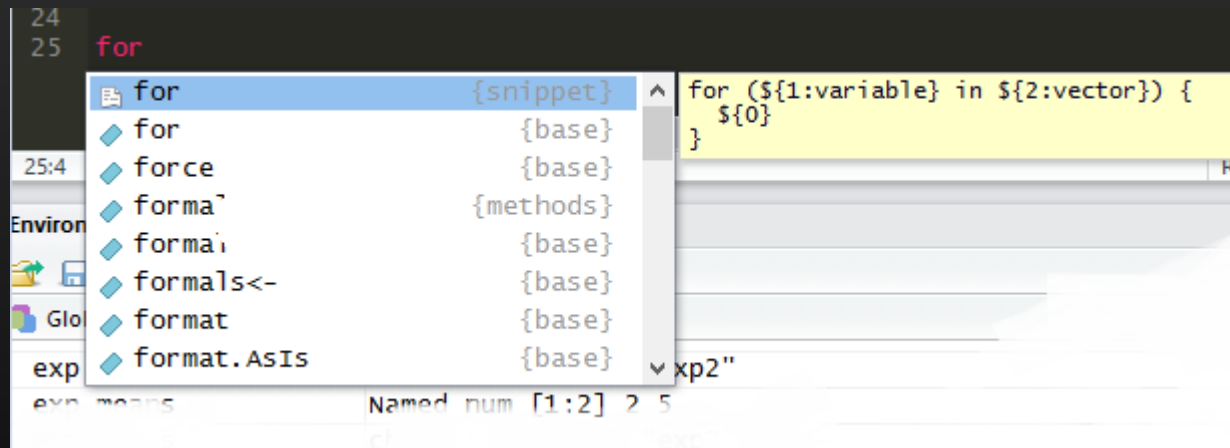
- Great built-in features:
  - code completion on tab



# RStudio helps again..

● Introduction   ● Project Set-Up   ● Data Management   ● Coding   ● Conclusion

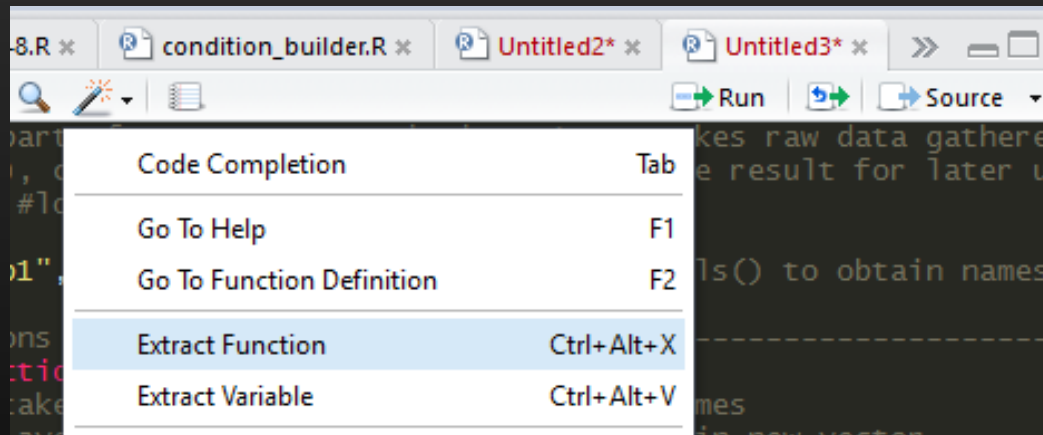
- Great built-in features:
  - code completion on tab
  - code snippets for loops / if-else / functions



# RStudio helps again..

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- Conclusion

- Great built-in features:
  - code completion on tab
  - code snippets for loops / if-else / functions
  - function wizard



# Conclusion

# Reproducible?

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---

- Using the provided tools and guidelines you should be able to:
  - understand some of the terminology used in books / online
  - find other helpful resources
  - develop your own workflow
  - move your analyses between computers
  - share your work with others and collaborate with them
  - **reproduce your research**



# Reproducible?

● Introduction   ● Project Set-Up   ● Data Management   ● Coding   ● Conclusion

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  - move your analyses between computers
  - share your work with others and collaborate with them
  - **reproduce your research**

... Enjoy.

Thanks for enduRing!