



# Reproducible projects in R: practical workshop

Marina Vabistsevits



# What this workshop is about

1. **Project-oriented workflow** (organising your projects with .Rproj and renv)
  - 1.1. Rstudio efficiency tips
2. Using **Git** to track changes in your R projects
3. **R projects on UKB-RAP/All of Us** platforms with Git



Icon for R



Icon for RStudio



software  
company who  
develop  
Rstudio

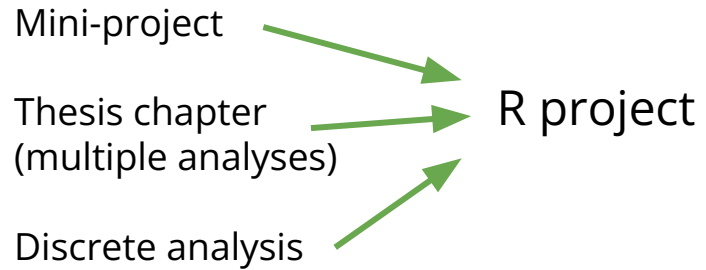


**Positron** -  
new IDE from  
Posit for  
R/Python etc

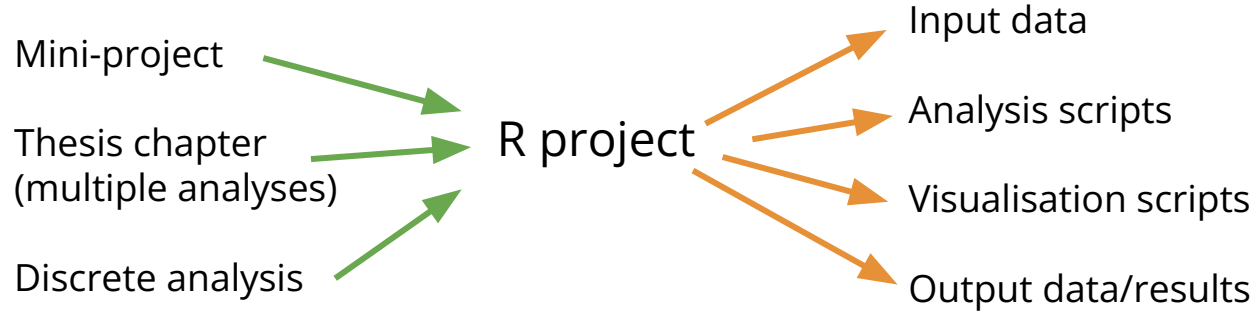
*The slides will be shared after!*

# 1. Project-oriented workflow

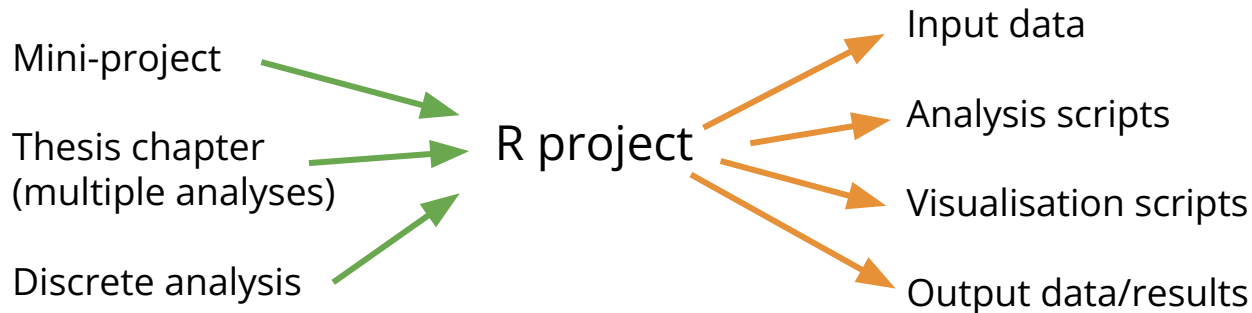
# R projects



# R projects



# R projects



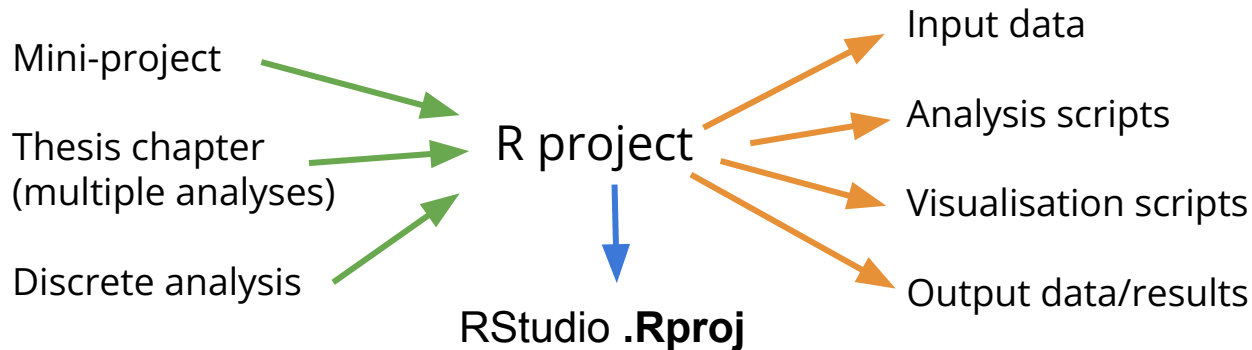
How you do organise your projects? scripts? folders?

Where do you store and access your data? results? figures?

Does it matter? Are your analysis/projects reproducible?

Can you organise your projects better and make life easier for future self/colleagues?

# R projects



How you do organise your projects? scripts? folders?

Where do you store and access your data? results? figures?

Does it matter? Are your analysis/projects reproducible?

Can you organise your projects better and make life easier for future self/colleagues?

Reproducibility can be enhanced through intentionally organising projects with .Rproj, i.e. working in **project-oriented workflow**

## Project-oriented workflow



Photo by secumem

📅 2017/12/12

👤 Jenny Bryan

I was honored to speak this week at the IASC-ARS/NZSA Conference, hosted by the Stats Department at The University of Auckland. One of the conference themes is to celebrate the accomplishments of Ross Ihaka, who got R started back in 1992, along with Robert Gentleman. My talk included advice on setting up your R life to maximize effectiveness and reduce frustration.

Two specific slides generated much discussion and consternation in #rstats Twitter:

*If the first line of your R script is*

```
setwd("C:\\Users\\jenny\\path\\that\\only\\I\\have")
```

*I will come into your office and SET YOUR COMPUTER ON FIRE 🔥.*

*If the first line of your R script is*

```
rm(list = ls())
```

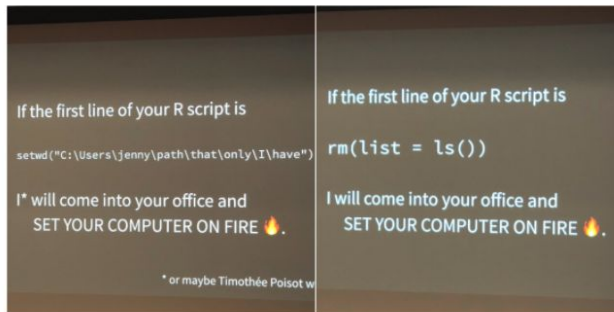
*I will come into your office and SET YOUR COMPUTER ON FIRE 🔥.*

Twitter profile card for Jenny Bryan (@JennyBryan). It includes her profile picture, a 'Following' button, and a bio: 'Software engineer @rstudio, humane #rstats, adjunct prof @UBC where I created @STAT545, part of @ropensci, she/her'. Location is 'Vancouver, BC' and website is 'jennybryan.org'.

**Hadley Wickham**   
@hadleywickham

Follow

The only two things that make @JennyBryan 🤔😡💩. Instead use projects + here::here() #rstats



4:50 PM - 10 Dec 2017

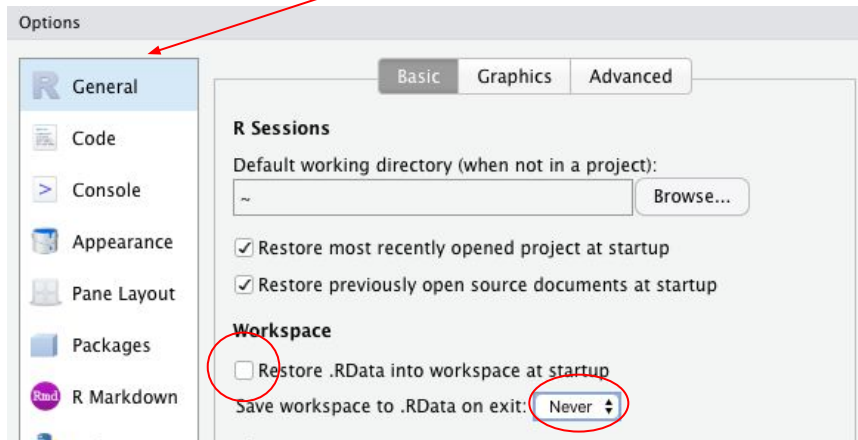
290 Retweets 950 Likes





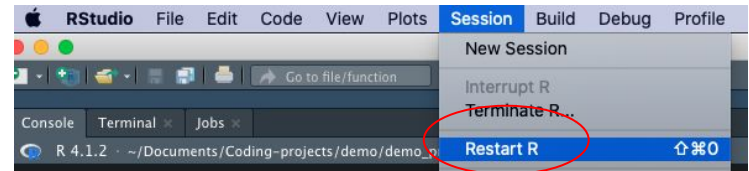
# Save your source code, not the workspace

**Do not** save your .Rdata workspace:  
(untick and select 'never')



**Do not use** `rm(list = ls())`

Restart R daily\* to ensure a clean environment:



Save important objects and intermediates to files and scripts, in a modular way

If there is no source, it's not reproducible

# Why setting directory does not work

Don't use `setwd()`

```
setwd("path/that/only/works/on/my/machine")
```

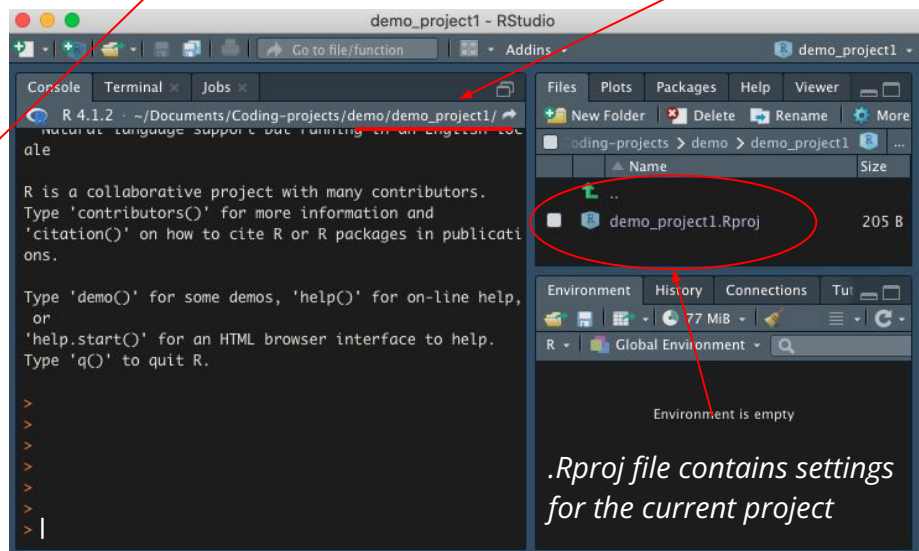
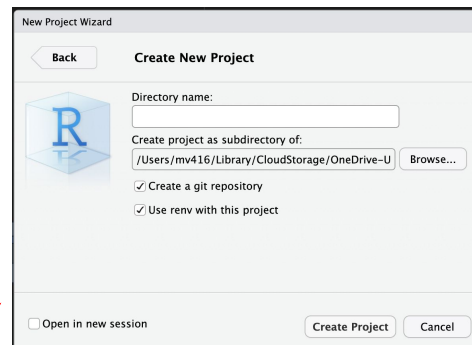
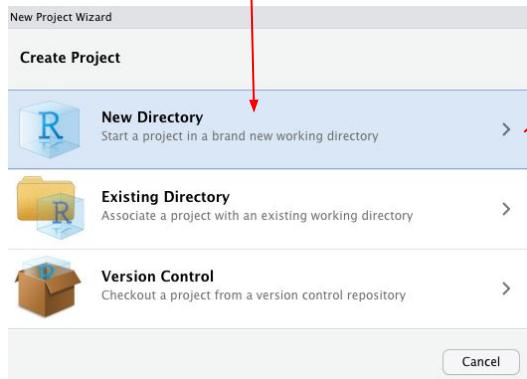
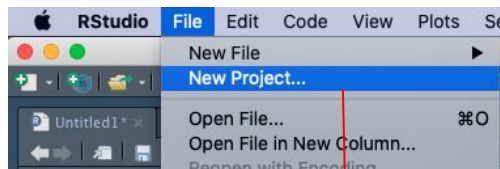
- Only works on your system
- Unlikely to work in a few years or on another computer
- Relying on hard-coded paths make a project easy to break and potentially not reproducible (shared data paths on the server is a different matter)

## The solution to both:

- keep your work as an `.Rproj` (i.e. project-oriented workflow) to help you manage your workspace env and file paths

# Project-oriented workflow

Use Rstudio / .Rproj for your data analysis projects



*Project directory*

*.Rproj file contains settings  
for the current project*

*This means that you are essentially compartmentalizing your current project*

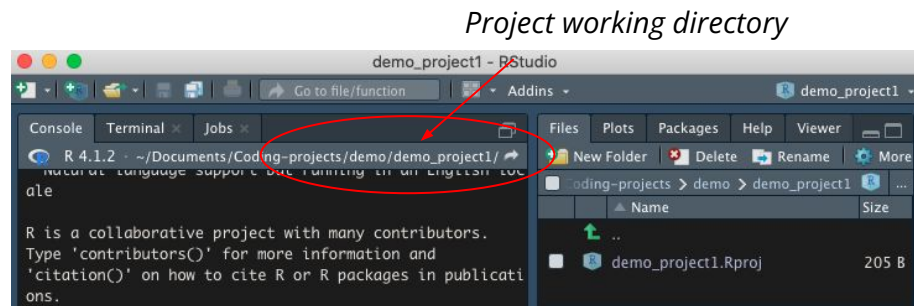
# Using .Rproj for your data analysis projects:

**File system discipline:** project directory stores all your data, scripts, figures

**File path discipline:** The working directory is set to the project directory (e.g.

**demo\_project1/**), so you don't need to *setwd()* or specify full paths to data (only internal subfolders that are relative to top directory)

**Working directory intentionality:** when working on project A, your working directory is set to project A's folder



- The project creates everything it needs, within its workspace/folder, and touches nothing it did not create
- Any scripts are written assuming they will be run from a fresh R session within the project
- The project folder can be moved *anywhere*, and everything will still work (no paths will be broken)

# You don't need to use setwd()

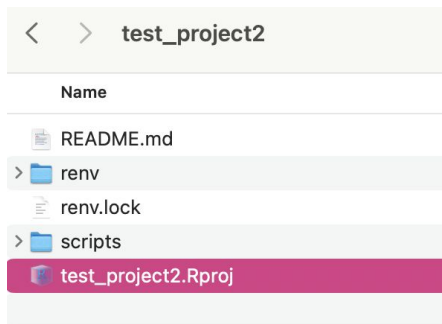
Keeping your work as an **.Rproj** will help you manage your file paths:

```
(base) [demo_project1] tree -L 3
├── README.md
├── data
│   ├── processed
│   │   └── data_file_clean.tsv
│   └── raw
│       └── raw_data_file.tsv
├── demo_project1.Rproj
├── figures
├── results
└── scripts
    ├── 01_load_raw_data_and_clean.R
    ├── 02_explore_data.R
    ├── 03_run_analysis.R
    ├── 04_make_plots.R
    └── 05_project_summary.Rmd
```

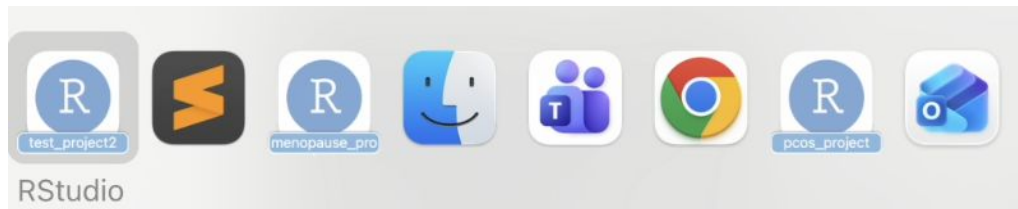
```
01_load_raw_data_and_clean.R
1 library(readr)
2
3 # read raw_data_file.tsv
4 data_raw <- read_tsv("data/raw/raw_data_file.tsv")
5
6 # clean data
7 # <cleaning code>
8 data_clean <- data_raw
9
10 # save cleaned data
11 write_tsv(data_clean, "data/processed/data_file_clean.tsv")
12
13
14
```

# You don't need to use `rm(list=ls())`

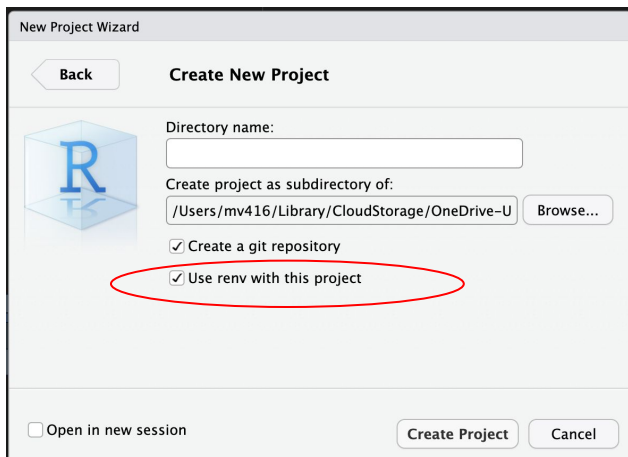
Click on .Rproj file in project directory to open a new clean session in your project:



You can also have multiple non-conflicting projects open at the same time:



# renv for managing package versions within projects



New Project Wizard

Back Create New Project

Directory name:

Create project as subdirectory of:

/Users/mv416/Library/CloudStorage/OneDrive-U Browse...

☒ Create a git repository

☒ Use renv with this project

☐ Open in new session

Create Project Cancel

**renv** package helps you create reproducible environments for your R projects



- Keeps track of all your installed packages
- Installs them in a new location (i.e. easy to share)
- Useful for working on UKB-RAP/AoU platforms (as a part of .Rproj)

All R packages  
installed on your  
computer

Packages for  
“phenotype A”  
.Rproj in renv

# Setting up your renv

```
install.packages("renv")
```





`renv::init()` – initialise your project env

This creates:

NB this is the same as

☒ Use renv with this project

- **renv/**: stores packages for the project.
- **renv.lock**: records packages and the exact versions used
- **.Rprofile**: ensures renv activates when the project opens

<input type="checkbox"/>		.Rprofile	26 B
<input type="checkbox"/>		renv	
<input type="checkbox"/>		renv.lock	402 B
<input type="checkbox"/>		test_project1.Rproj	205 B

When renv is initialised, your project gets its own project library - a private folder (**renv/library/**) where packages are installed just for this project.

This means package versions are isolated and recorded for each project, so updating a package in one project won't break others.

This makes collaboration and reproducibility much easier.

**renv::snapshot()** – update your env with newly installed packages (saves to renv.lock)

**renv::restore()** – when project is re-opened, load all your required packages

- should not be necessary to run in the local Rstudio, only when cloning a project from Git (eg on UKB-RAP)

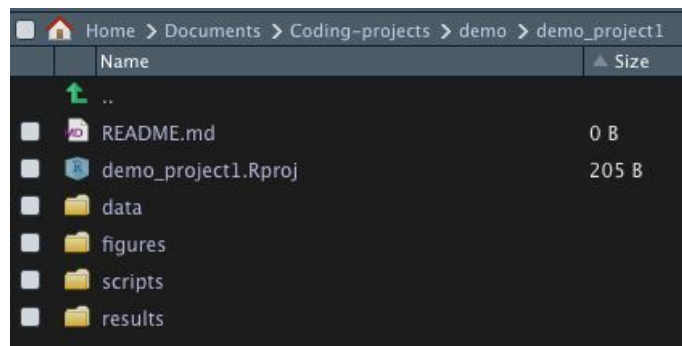
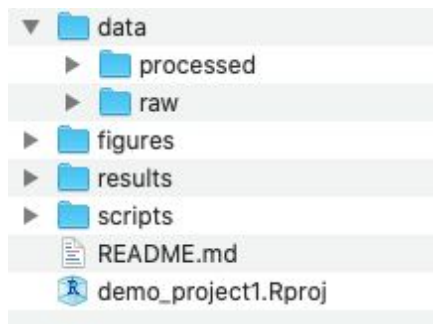


# Practical part (15 mins)

1. In your local Rstudio, create a new Project (e.g. *test\_project1*) with **renv** and **git** enabled in a new session (window)
  - a. Create a new R script in the project folder; install package “sp” in it
  - b. Check the “sp” package loads in your project, then *renv::snapshot()* it
  - c. Check the “sp” package loads in your other ongoing open Rstudio session
  - d. Close your new project Rstudio window
  - e. Re-open by clicking on *test\_project1.Rproj* file
  - f. Test the “sp” package still loads (do not install again!)
2. Download ‘demo\_project1.zip’ from the email and uncompress it
  - a. Open the project and explore the folders/files
  - b. Has the project been created with *renv*?
  - c. Review *00\_set\_up\_renv.R*; restore the env and install a package a new package (eg “sp”)

## 1.1. Hands-on tips for your R projects

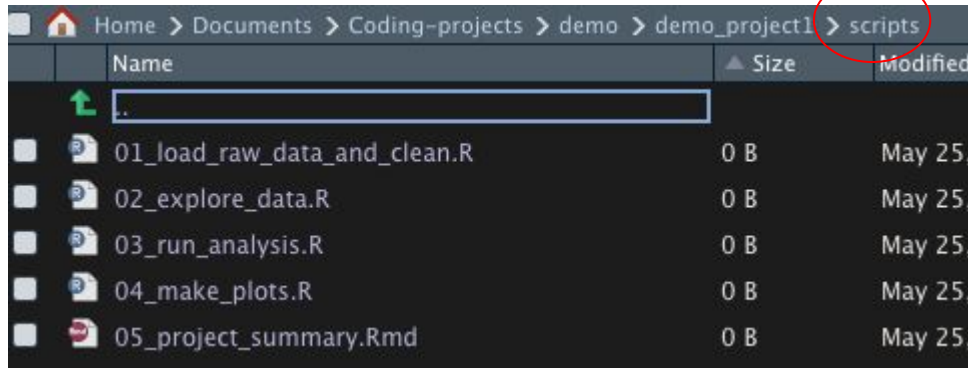
# Organise your projects intentionally



# Take advantage of default ordering

- 01\_load\_raw\_data\_and\_clean.R
- 02\_explore\_data.R
- 03\_run\_analysis.R
- 04\_make\_plots.R
- 05\_project\_summary.Rmd

Can have many parts of the analysis separately - save interim results as files and re-read then in the next script

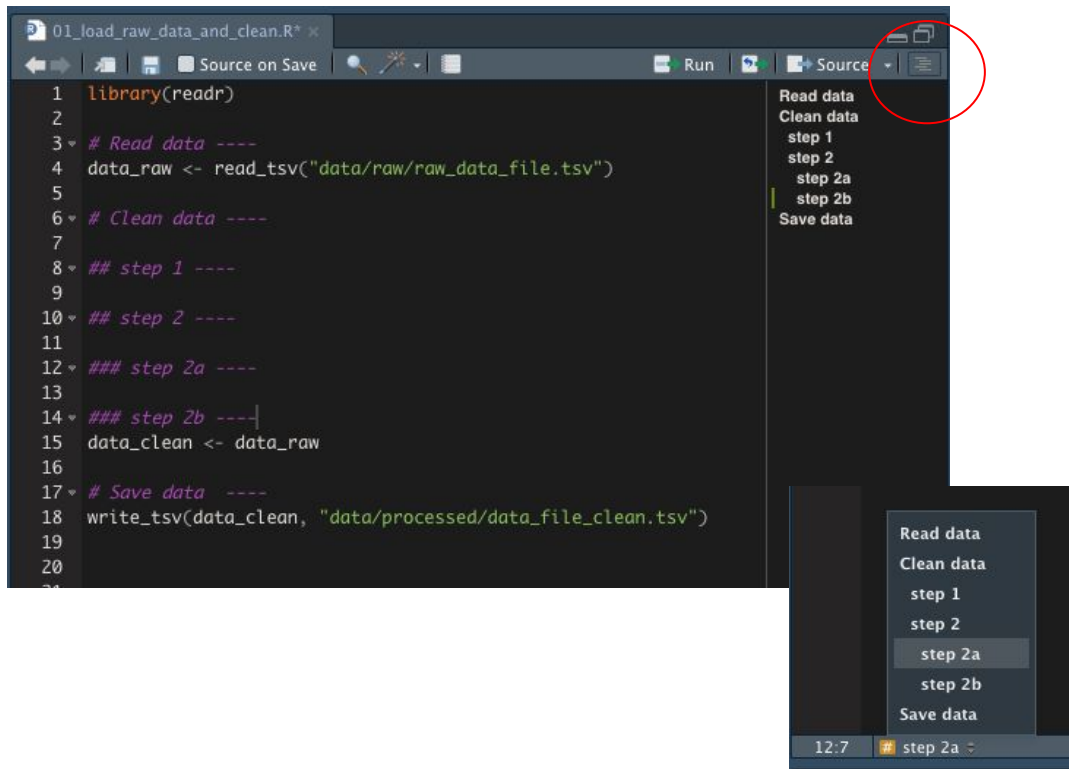


Home > Documents > Coding-projects > demo > demo\_project1 > scripts

	Name	Size	Modified
↑	.		
📄	01_load_raw_data_and_clean.R	0 B	May 25,
📄	02_explore_data.R	0 B	May 25,
📄	03_run_analysis.R	0 B	May 25,
📄	04_make_plots.R	0 B	May 25,
📄	05_project_summary.Rmd	0 B	May 25,

“How to name files” by  
Jenny Bryan - [link](#) (5 mins)

# Name your code sections and use them for quick navigation



- Use section headings:

# section ----

## subsection ----

### subsubsection ----

- Great for navigating in long scripts
- Can fold sections

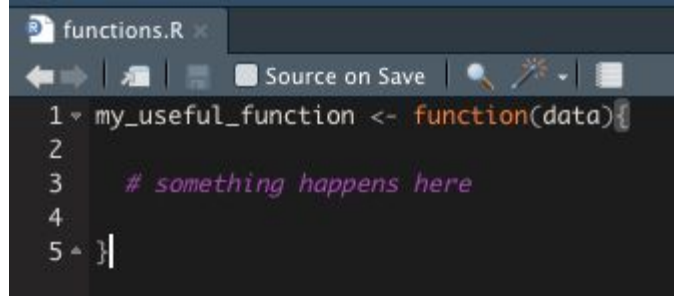


# Jump to function definition or open data frame

```
## step 2 ----  
  
output <- my_useful_function(input)
```

To navigate to a function definition or review a function from a package:

- **Cmd + click** on the function name

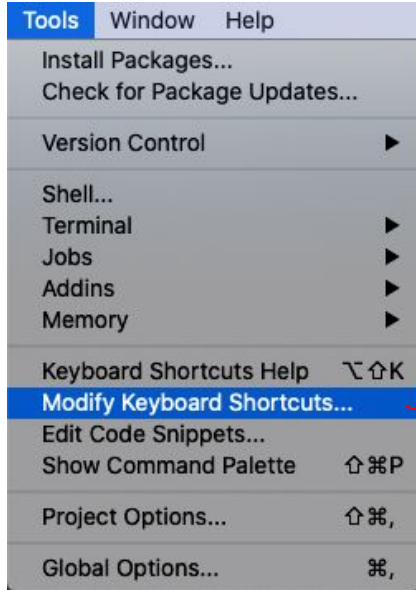




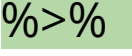
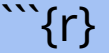

The screenshot shows an RStudio editor window with a tab labeled 'functions.R'. The code in the editor is as follows:

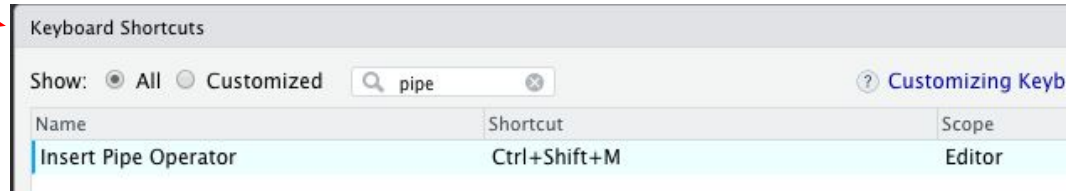
```
1 my_useful_function <- function(data){  
2  
3   # something happens here  
4  
5 }
```

> to see what it does internally /  
check arguments

# Keyboard shortcuts



-  (option + Enter)
  -  (option/alt + " - ")
  -  (control + shift + M)
  -  (control + shift + I)
- 

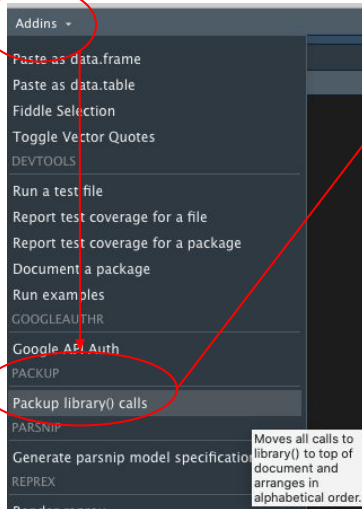


# Move all libraries to the top

Install *packup* add-in:

```
devtools::install_github("milesmbain/packup")
```

Call it from Addins menu:



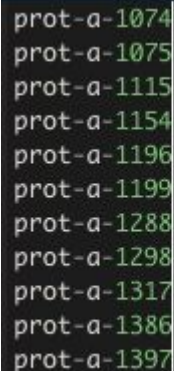
```
functions.R 03_run_analysis.R* 02_explore_data.  
Source on Save  
1 library(readr)  
2 data_clean <- read_tsv("data/processed/data_file_clean.tsv")  
3  
4  
5  
6  
7  
8  
9 # some analysis here  
10 library(TwoSampleMR)  
11  
12  
13  
14  
15  
16  
17 # other analysis here  
18 library(tidyr)  
19  
20  
21  
22  
23  
24  
25  
26 # maybe some viz here  
27 library(ggplot2)  
28
```

```
functions.R 03_run_analysis.R* 02_explore_data.R 01_load_  
Source on Save  
1 library(ggplot2)  
2 library(readr)  
3 library(tidyr)  
4 library(TwoSampleMR)  
5  
6 data_clean <- read_tsv("data/processed/data_file_clean.tsv")  
7  
8  
9  
10  
11  
12  
13 # some analysis here  
14  
15  
16  
17  
18  
19  
20  
21 # other analysis here  
22  
23  
24  
25  
26  
27  
28  
29  
30 # maybe some viz here  
31
```



# Vertical selection

(hold *option* or *alt* and drag cursor down to select vertically)



```
prot-a-1074  
prot-a-1075  
prot-a-1115  
prot-a-1154  
prot-a-1196  
prot-a-1199  
prot-a-1288  
prot-a-1298  
prot-a-1317  
prot-a-1386  
prot-a-1397
```

A screenshot of a code editor with a dark background. It shows a list of identifiers, each on a new line: 'prot-a-1074', 'prot-a-1075', 'prot-a-1115', 'prot-a-1154', 'prot-a-1196', 'prot-a-1199', 'prot-a-1288', 'prot-a-1298', 'prot-a-1317', 'prot-a-1386', and 'prot-a-1397'. A vertical selection is active, indicated by a thin white line and a light blue highlight, covering the ID numbers (1074, 1075, 1115, 1154, 1196, 1199, 1288, 1298, 1317, 1386, 1397) in the last column of each line.

Great for e.g.

- commenting out a block of code with #
- adding " " around a column of ids

Any other R tips to share?

## 2. Using Git to track changes in your R projects

*<https://happygitwithr.com/>*

# Checklist before we start

- GitHub account
- Personal access token
- Git is available on your laptop
- SSH-key (linking Git on your laptop with your GitHub account)

# Git brief intro

**Git** - tool for code version control

- Tracking code changes
- Keeping older versions of the script
- Coding collaboration (with others or yourself in multiple locations)
- Tracking who and when made changes

} If used consistently!

## Why use it?

- Can facilitate data/code integrity
- Improves reproducibility (e.g. keeps record of changes)
- Enables code sharing with colleagues / as a part of publication
- Important skill for anyone working with data (e.g. can showcase your work during job applications)

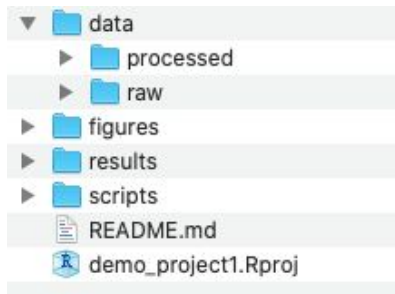
## Github / Gitlab etc

- Platforms where you can store your coding projects using Git

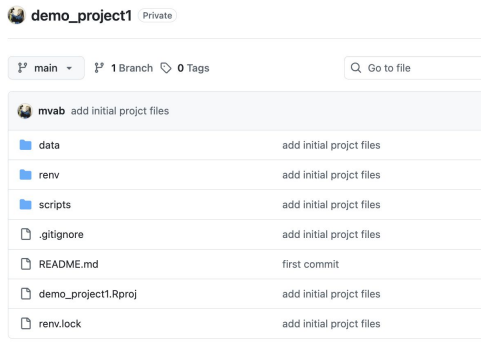


# Why might you want to use git? (scenario 1)

Your R project  
on local Mac

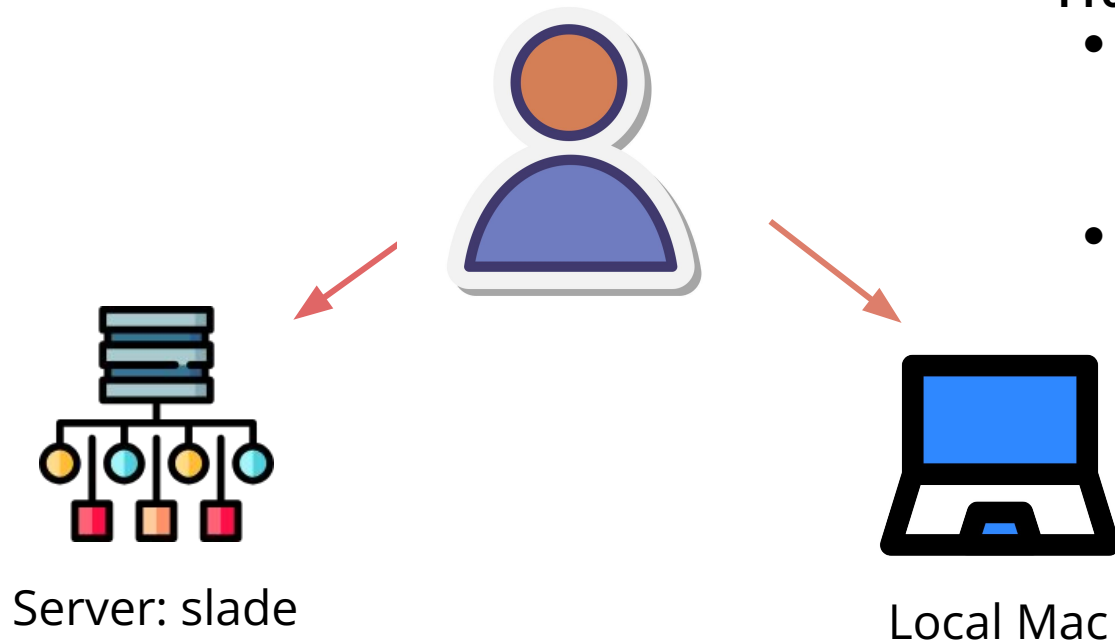


Your R project  
as GitHub repo



- Using git helps you keep track of your daily changes
- All your work is backed-up (as long as you do it regularly and intentionally)
- “Collaborating with yourself” in one place

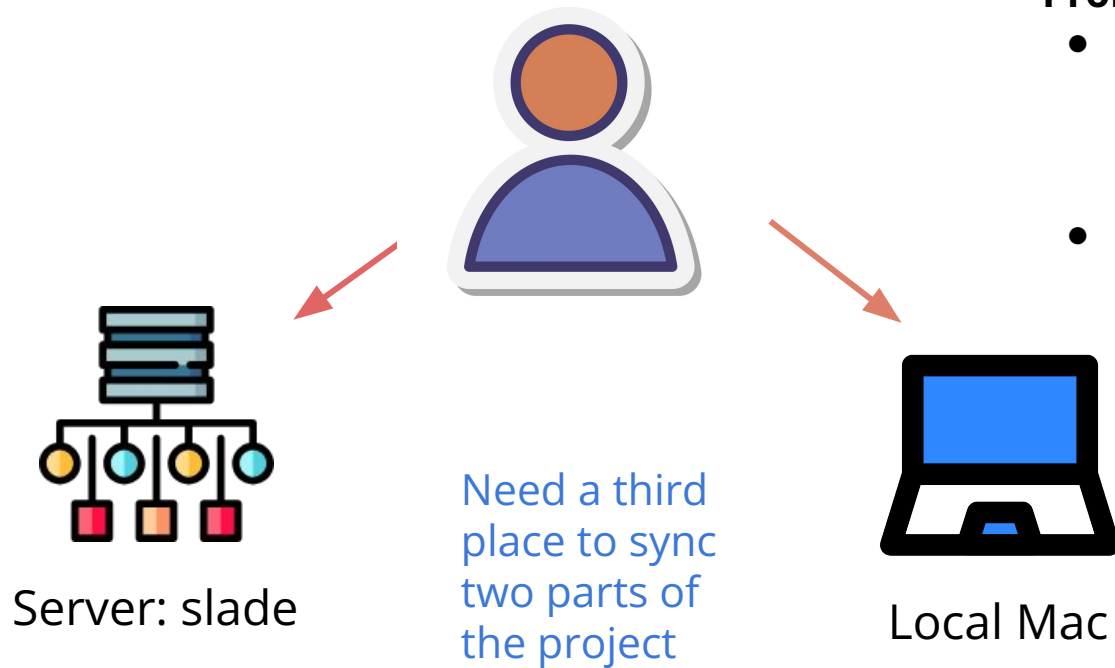
# Why might you want to use git? (scenario 2)



## Problem:

- The same project is both locations (e.g due to data restrictions / tool availability)
- Don't want to copy over scripts all the time, as it might get messy!

# Why might you want to use git? (scenario 2)

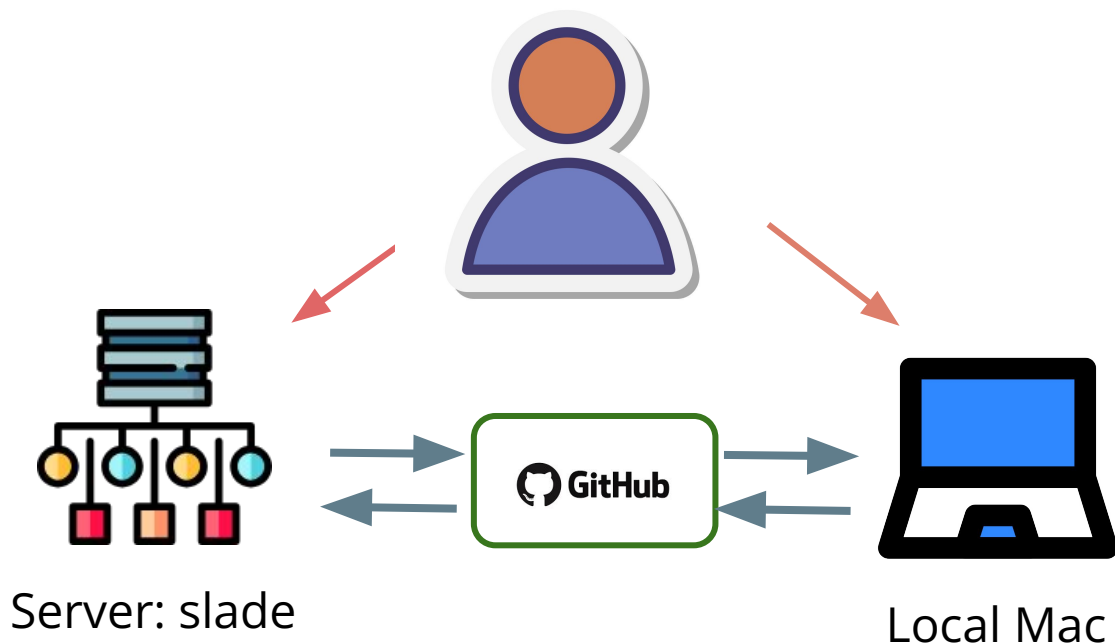


## Problem:

- The same project is both locations (e.g due to data restrictions / tool availability)
- Don't want to copy over scripts all the time, as it might get messy!



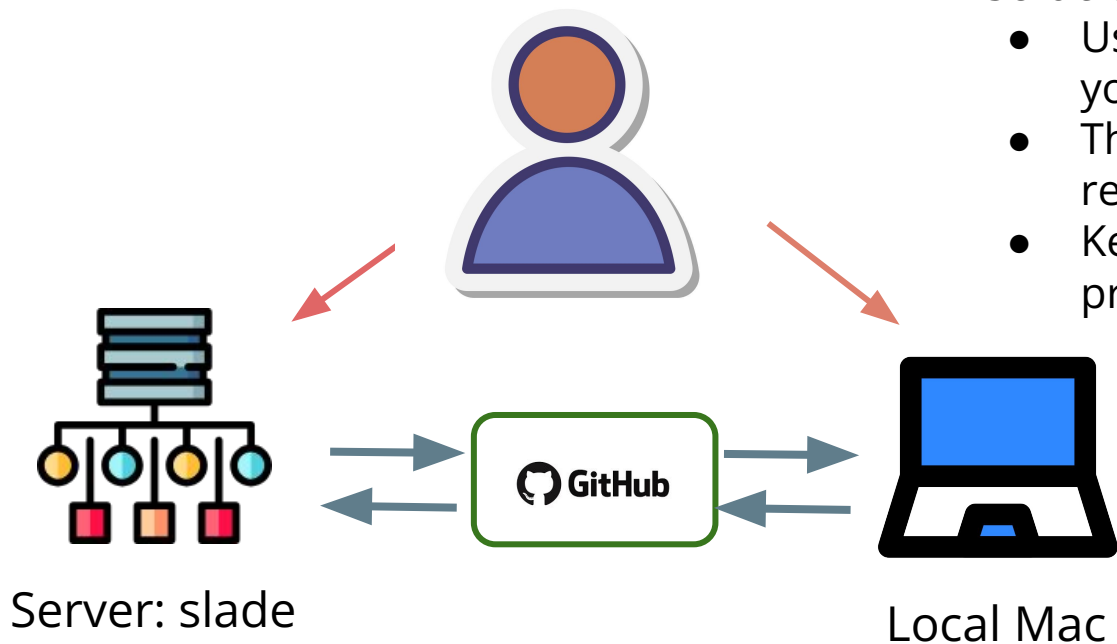
# Why might you want to use git? (scenario 2)



# Why might you want to use git? (scenario 2)

## Solution:

- Use git to “collaborate with yourself” in two places
- This enhances project reproducibility
- Keeps all analysis code in one project folder/git repository



# Why might you want to use git? (scenario 3)

Your R project  
on

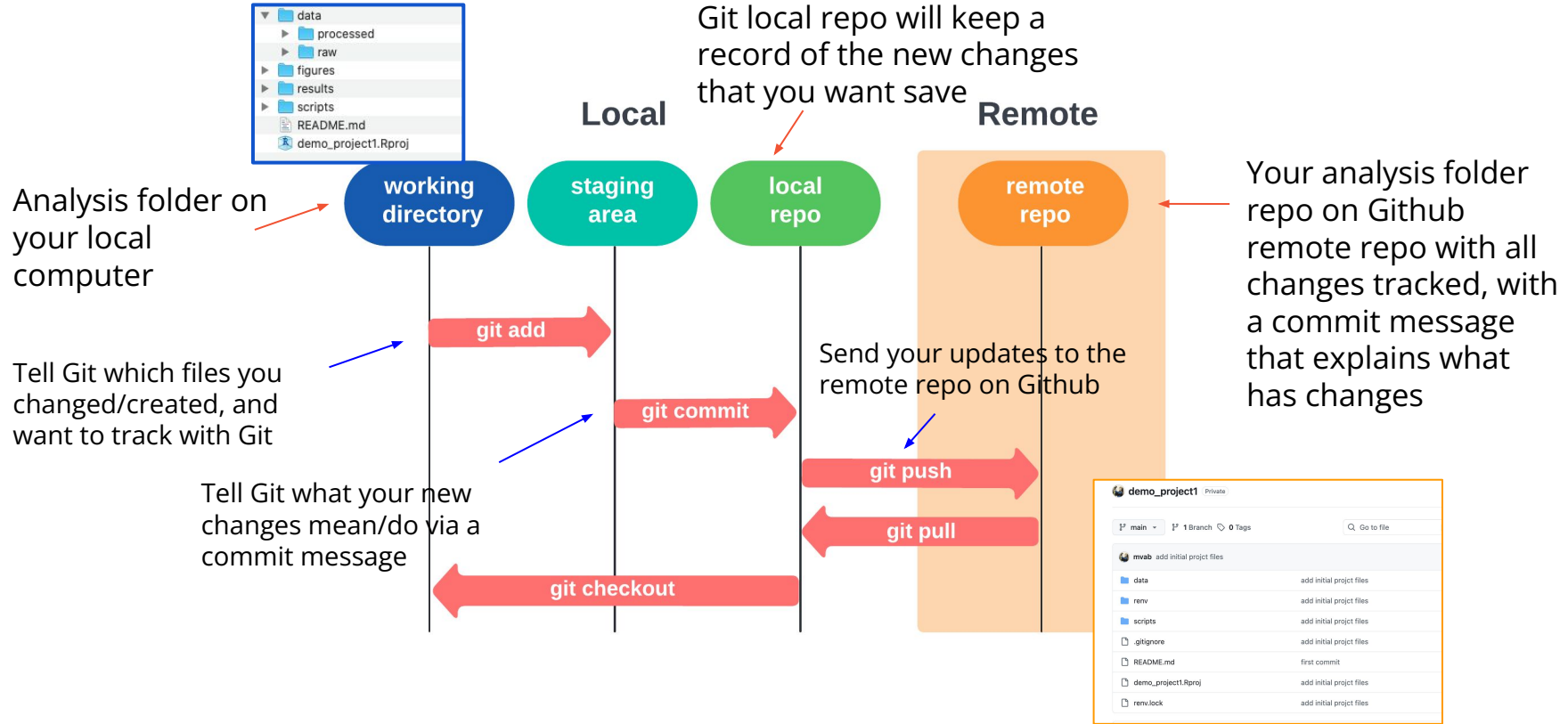


Your R project  
as GitHub repo



- Using git helps you keep track of your daily changes
- All your work is backed-up (you are 'forced' to use Git daily to back up your work, because you start a new R instance every time you work on your project)

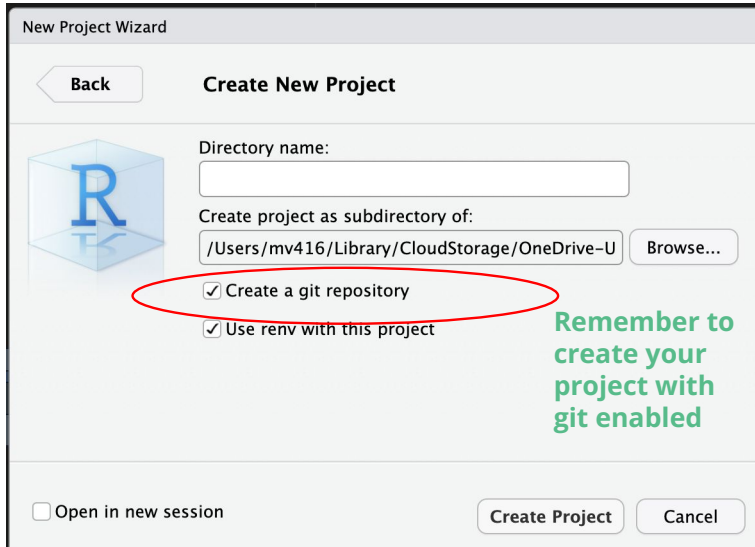
# Basic git commands / actions



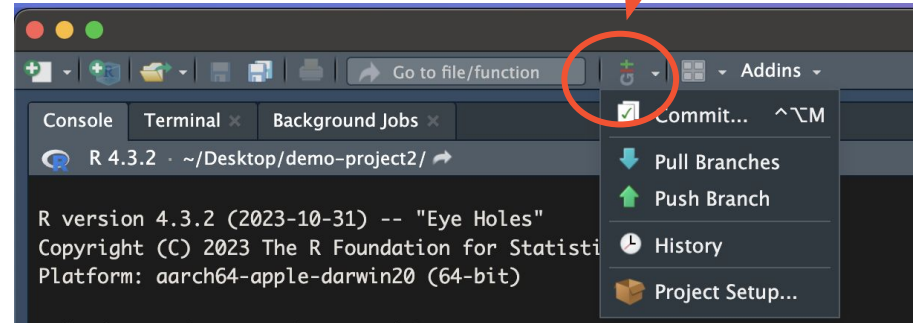
Follow along if you can

# Using git with your .Rproj

- Rstudio user interface makes it easy to get started with git
- (but also can use the terminal to run git commands)



This means Git is enabled in the current project (use drop-down to interact with it)



-> Next need to link your local project folder to a GitHub repo

# Create remote repo

After logging in your Github account:

To create a new repository:

**Repositories -> New**

## Create a new repository

Repositories contain a project's files and version history. Have a project elsewhere? [Import a repository](#).  
Required fields are marked with an asterisk (\*).

### 1 General

Owner \*



Repository name \*

test\_project1

test\_project1 is available.

Give repo the same name as your .Rproj folder

Great repository names are short and memorable. How about [legendary-memory](#)?

Description

0 / 350 characters

Make it private if it's ongoing work

### 2 Configuration

Don't change any configurations to keep it simple (except visibility)

Choose visibility \*

Choose who can see and commit to this repository

Public

Start with a template

Templates pre-configure your repository with files.

No template

Add README

READMEs can be used as longer descriptions. [About READMEs](#)

Off

Add .gitignore

.gitignore tells git which files not to track. [About ignoring files](#)

No .gitignore

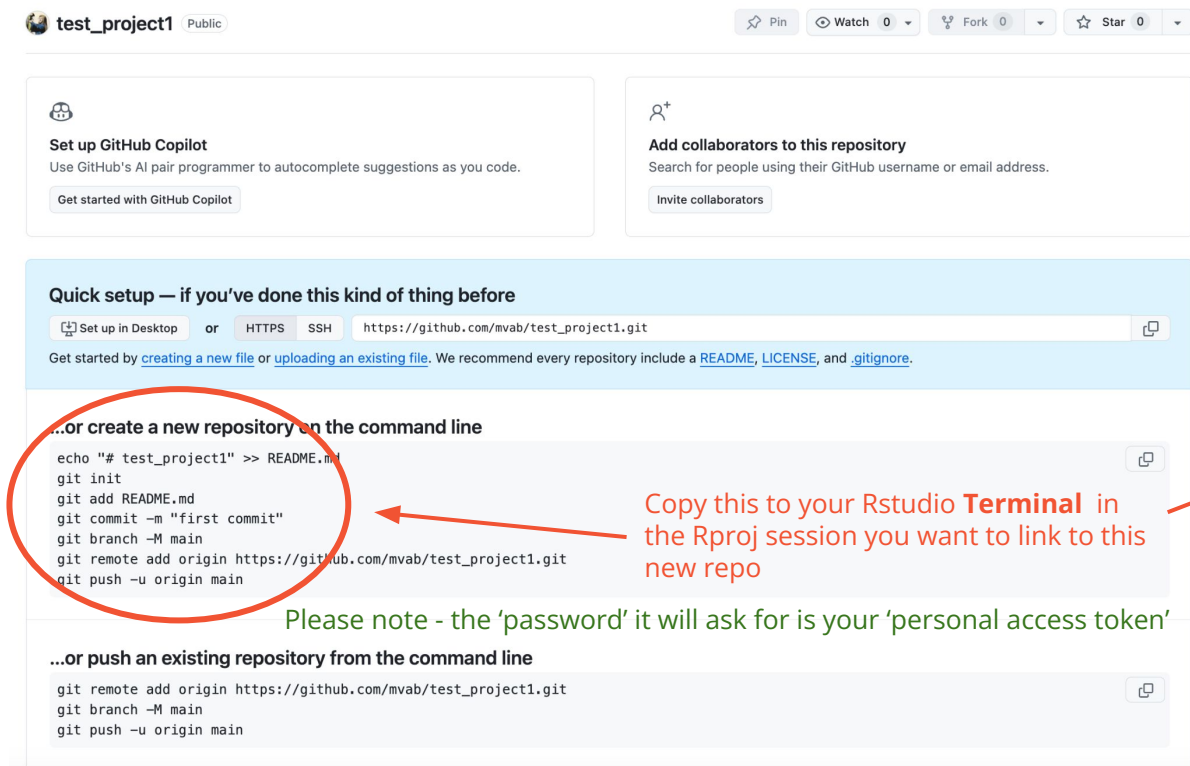
Add license

Licenses explain how others can use your code. [About licenses](#)

No license

Create repository

# GitHub makes it easy to link your .Rproj to the new repo:



The screenshot shows the GitHub repository page for 'test\_project1'. At the top, there are buttons for 'Pin', 'Watch' (0), 'Fork' (0), and 'Star' (0). Below these are two cards: 'Set up GitHub Copilot' and 'Add collaborators to this repository'. The main section is titled 'Quick setup — if you've done this kind of thing before' and shows the repository URL 'https://github.com/mvab/test\_project1.git'. Below this, there are two sections for creating a new repository or pushing an existing one from the command line. The first section, '...or create a new repository on the command line', is circled in red and contains the following commands:

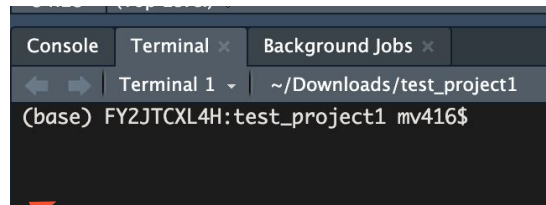
```
echo "# test_project1" >> README.md
git init
git add README.md
git commit -m "first commit"
git branch -M main
git remote add origin https://github.com/mvab/test_project1.git
git push -u origin main
```

The second section, '...or push an existing repository from the command line', contains the following commands:

```
git remote add origin https://github.com/mvab/test_project1.git
git branch -M main
git push -u origin main
```

Copy this to your Rstudio **Terminal** in the Rproj session you want to link to this new repo

Please note - the 'password' it will ask for is your 'personal access token'

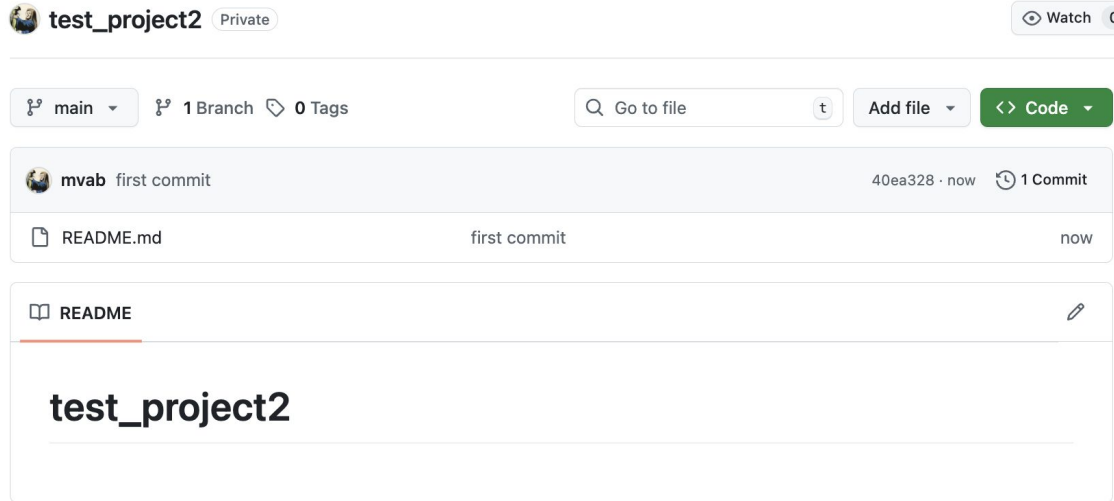


The screenshot shows an RStudio Terminal window with the following content:

```
Console Terminal x Background Jobs x
Terminal 1 ~ /Downloads/test_project1
(base) FY2JTCXL4H:test_project1 mv416$
```

An orange arrow points from the text 'Copy this to your Rstudio Terminal in the Rproj session you want to link to this new repo' to the Terminal window.

# Remote repo after first commit

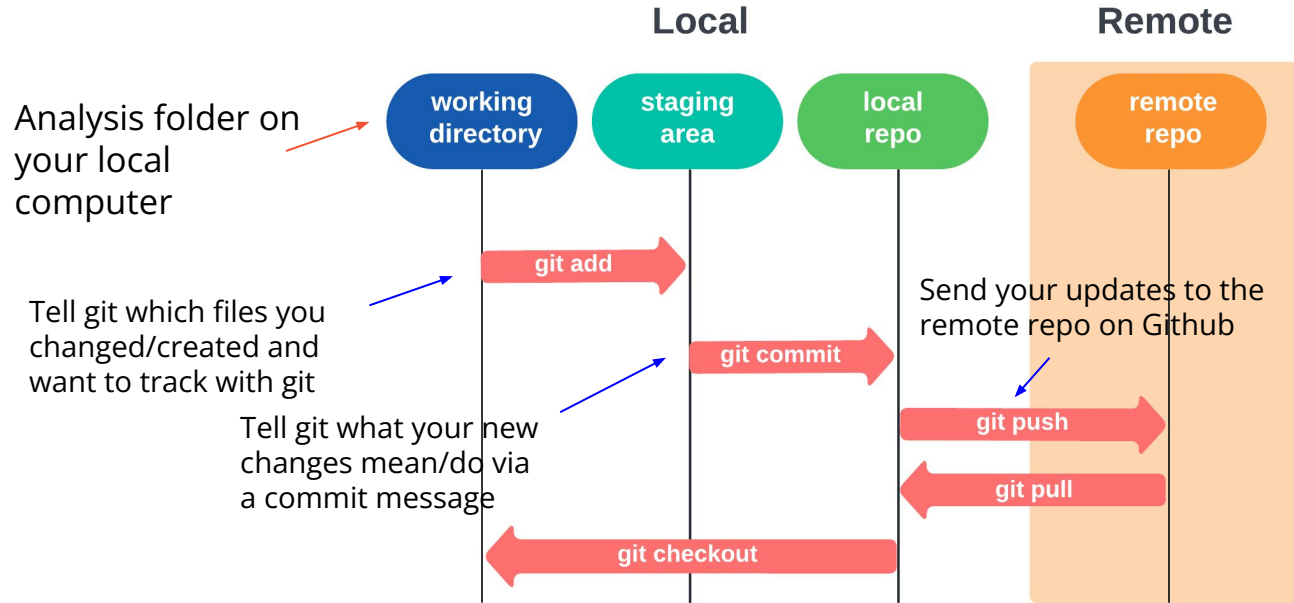


Only  
README file  
has been  
added



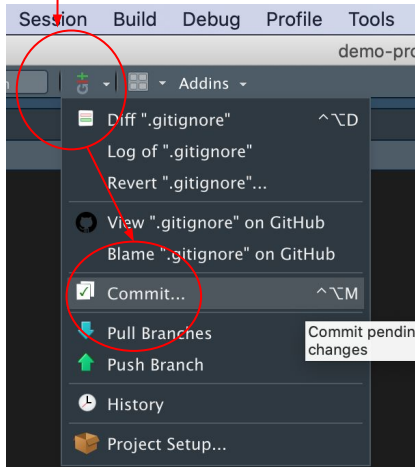
```
echo "# test_project1" >> README.md
git init <- initialises your project repo
git add README.md
git commit -m "first commit"
git branch -M main
git remote add origin https://github.com/mvab/test_project1.git
git push -u origin main
```

} initialises your project repo

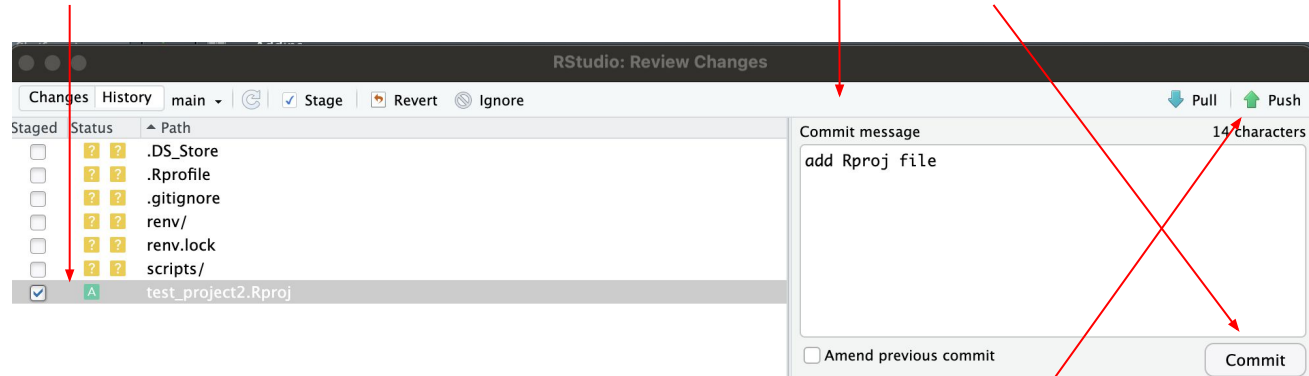


# Time to add other files!

1) Git button - > Commit




2) Select the files you want Git to **add** (track):







3) write your 'commit message', ie description of your change (tip: write it as an instruction), then press "**Commit**"


4) Press "**Push**" to send the change to remote repo


# Your change has been added:



 **test\_project2** Private


 Watch 0

 main  1 Branch  0 Tags

 Add file


 Code

 **mvab** add Rproj file 85cc036 · 18 minutes ago  2 Commits

 README.md


first commit

37 minutes ago

 test\_project2.Rproj

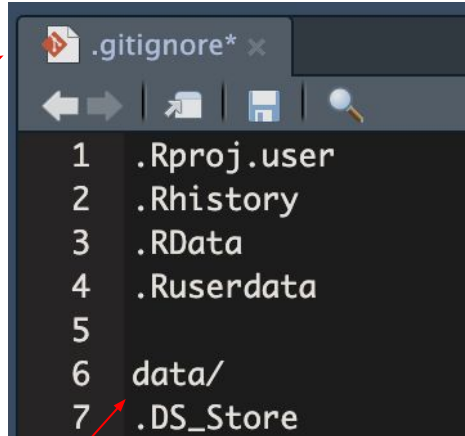
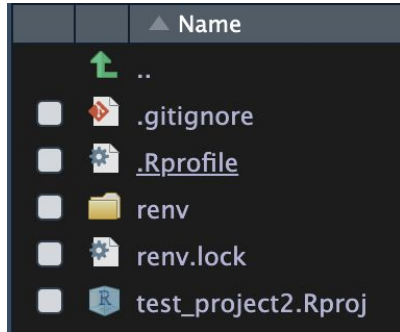
add Rproj file

18 minutes ago

 README

## test\_project2

# .gitignore file - list files you don't want to store in remote repo



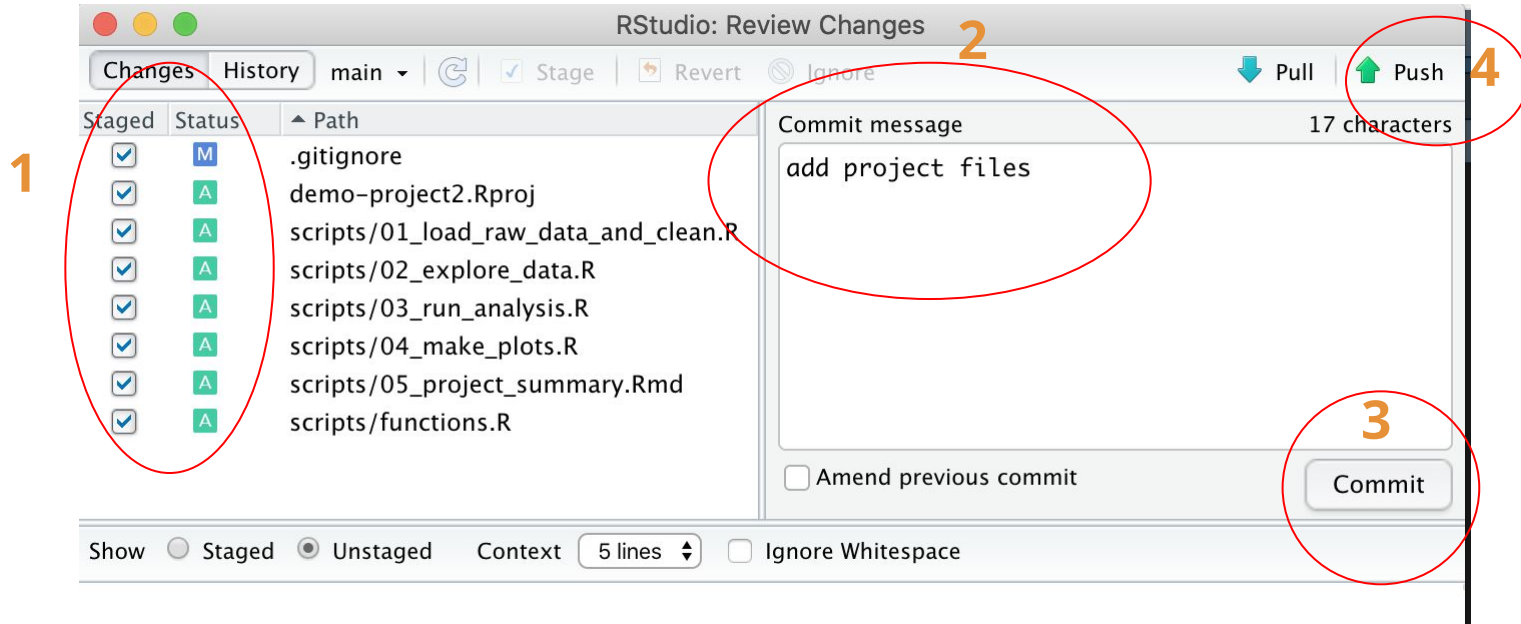
Add **data/** folder to *.gitignore* file so that your data files (if large or sensitive) are not committed to your project repo on Github

- list files and folders in your project folder in the *.gitignore* file - they will be ignore by Git (this prevents them from being accidentally added to GitHub repo)
- save and '**add**' file
- **commit** and **push** the updated *.gitignore* file to your repo

# Now add more files:

Commit changes:

Add message:



 main  1 Branch  0 Tags

 Go to file








t

Add file

 Code

 **mvab** add all scripts

6da936d · now  5 Commits

 renv	add renv files	now
 scripts	add all scripts	now
 .Rprofile	add renv files	now
 .gitignore	add gitignore	1 minute ago
 README.md	first commit	3 days ago
 renv.lock	add renv files	now
 test_project2.Rproj	add Rproj file	3 days ago

 README



# test\_project2

Your changes  
on Github:

# Practise adding a specific change to a script:

1

```
01_load_raw_data_and_clean.R* x
Source on Save
1 library(readr)
2
3 # Read data ----
4 data_raw <- read_tsv("data/raw/raw_data_file.tsv")
5
6 # Clean data ----
7
8 ## step 1 ----
9
10 c = 1 + 2 # adding this change
11
12
13 ## step 2 ----
14
15 output <- my_useful_function(input)
16
17 ### step 2a ----
```

2

3

4

You will see what's changed here

Write a message that describes your change, then commit and push

```
RStudio: Review Changes
Changes History main Stage Revert Ignore Pull Push
Staged Status Path
[M] scripts/01_load_raw_data_and_clean.R
Commit message 12 characters
calculated c
[ ] Amend previous commit Commit
Show Staged Unstaged Context 5 lines Ignore Whitespace Unstage All
@@ -5,10 +5,13 @@ data_raw <- read_tsv("data/raw/raw_data_file.tsv")
5 5
6 6 # Clean data ----
7 7
8 8 ## step 1 ----
9 9
10 10 c = 1 + 2 # adding this change
11 11
12 12
10 13 ## step 2 ----
11 14
12 15 output <- my_useful_function(input)
13 16
14 17 ### step 2a ----
```

main ▾

demo-project2 / scripts /



mvab calculated c

..



01\_load\_raw\_data\_and\_clean.R

calculated c



02\_explore\_data.R

add project files



03\_run\_analysis.R

add project files



04\_make\_plots.R

add project files



05\_project\_summary.Rmd

add project files



functions.R

add project files



# Good practice principles

1. **Add** and **commit** your changes often and in small chunks
  - E.g. separate commits for different files and sections in your analysis steps
2. Write descriptive commit messages of what you've changed
  - "fix typos" is ok, but "update" is not
3. Use .gitignore
4. Keep a project README
  - describes what your scripts do, and where external data is stored

### 3. R projects on UKB-RAP/All of Us platforms with Git

*Start two new Rstudio instances now!*

# Working in Rstudio in on-demand instances

## The problem:

- You are forced to scratch from a clean env every time you work a project:
  - All packages need to be installed every time
  - All intermediate data tables need to be regenerated
  - To save progress, you have to copy your R/Rmd scripts back and forth with your User project



## The solution:

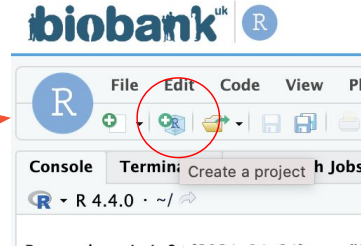
- Organise your R code as *Rproj* with *renv*, and use *Git* to back it up:
  - *renv::restore()* brings back all your packages
  - Intentional and modular file organisation helps you go back your intermediate files quicker
  - Committing your ongoing changes to your project repo on GitHub keeps your work backed up
  - *git clone* brings back your entire project to your clean Rstudio instance

Follow along if you can

# You already know how to make it work

Start a new project in Rstudio instance  
(instance #1 you started):

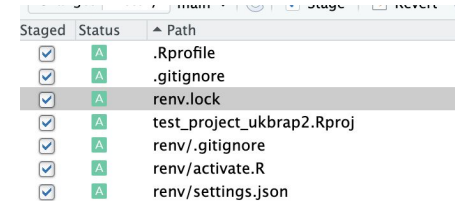
- 1) Create a new project (with renv + git)
- 2) On Github, create a new private repo
- 3) On Rstudio Terminal, authenticate:
  - `git config --global user.email "you@example.com"`
  - `git config --global user.name "user"`
- 4) Run the set up commands from GitHub on Rstudio Terminal
  - use your token as password!
- 5) Push all other files to your remote repo



...or create a new repository on the command line

```
echo "# test_project_ukbrap" >> README.md
git init
git add README.md
git commit -m "first commit"
git branch -M main
git remote add origin https://github.com/mvab/test_project_ukbrap.git
git push -u origin main
```

- 6) Upload or create `00_renv_setup.R` file project folder  
(attached in the email)
- 7) Add any other dummy R files you want



# renv setup

```
00_renv_setup.R x
Source on Save Run Source
1
2 # Project env initial set up ----
3
4 # only run at first use, or when installing additional packages for renv
5 lapply(c("dplyr", "readr"),
6       function(pkg) { if(! pkg %in% installed.packages()) {
7         install.packages(pkg)} } )
8
9 renv::snapshot() # run this after installing all packages (for the first time)
10
11 #####
12
13
14 # Return to project after cloning it to a new instance ----
15 renv::activate()
16 renv::restore()
17 library(dplyr) # test package load (othe rpackages can be loaded in the scripts they are used in)
18
19 # authenticate
20 system("git config --global user.email marina.vabi@hotmail.com")
21 system("git config --global user.name mvab")
22
```

1) Run once, only on the first package installation only + `renv::snapshot()` to save it to `renv.lock`

2) Then, push the new file and the updated `renv.lock` to GitHub:

Staged	Status	Path
<input checked="" type="checkbox"/>	A	00_renv_setup.R
<input checked="" type="checkbox"/>	M	renv.lock

3) Next time, when going back to this project, just need to activate and restore the renv

*Tip: keep a copy of 00\_renv\_setup.R file in all projects*

# Ending your R instance working on Rproj

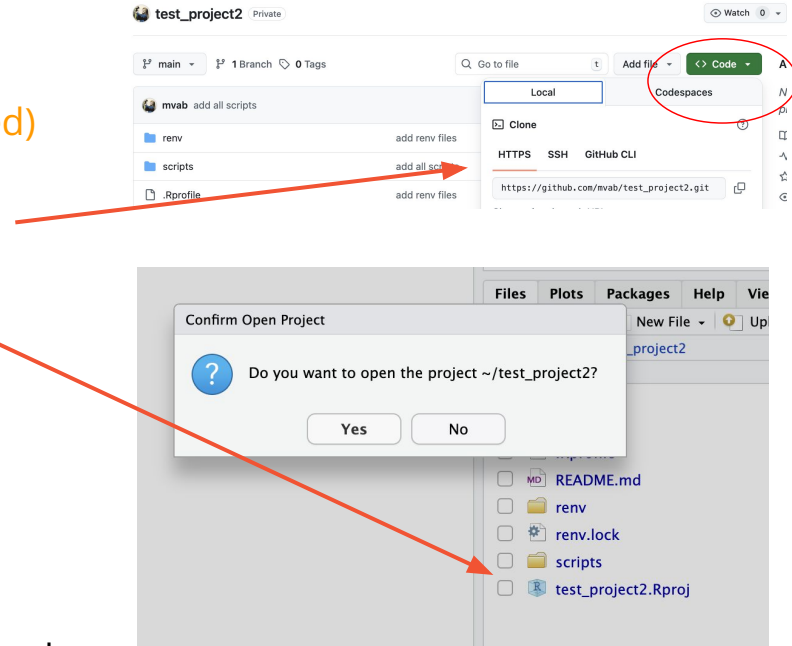
At the end of your session, save and push all your changes to GitHub (check it's all there)

- Remember to run `renv::snapshot()` - so that the `renv.lock` file is updated (push it to GitHub too!)
- Be especially careful if working with individual level data (tip: only save your data files in to folder `data/`, and list that folder in your `.gitignore` - so it does not get added to GitHub by accident)
- Once all code you wrote in the current session is on GitHub, it's safe to terminate  
instance #1

# Returning to your R project

- 1) Starting from a clean instance (instance #2 you started)
- 2) On Rstudio terminal, *git clone* your project
  - `git clone https://github.com/mvab/test\_project2.git`
  - Authenticate with your token
- 3) **Important:** click on Rproj file in the project folder to activate the project
- 4) Open `00_renv_setup.R` file and test that `library(dplyr)` currently does not work!
- 5) In `00_renv_setup.R` run:
  - `renv::activate()`
  - `renv::restore()`
  - Test dplyr again!

Any other packages you install - remember to run `renv::snapshot()` - so that the `renv.lock` file is updated (and push to your GitHub repo)







# Summary: using .Rproj for organising work

- “Reproducible project in R” means:
  - **File system discipline:** all files related to a single project are stored in a designated folder;
  - **Working directory discipline:** intentionally work in project directory when opening Rproj
  - **File path discipline:** paths are relative to the project directory (not hard-coded full paths!)
  - **Daily work habit:** Restarting R often and re-running your script under development from the top will help you catch issues early on
  - **Using git for version control**
- Practising these habits together will give you the biggest pay-off
  - Reproducing your analyses will be easy
  - Organising your projects will help you make sense of them in 6/12/etc months
  - Can move your project anywhere or share it with anyone without changing paths

# Final thoughts / disclaimers

- Project-oriented workflow in R is not suitable/applicable to every scenario
  - Sometimes data is stored externally, and can't be moved/copied (so you can't use within-project paths, but potentially can use symbolic links)
- Not all work is done interactively in Rstudio
  - Some people use R from the terminal on the server - again, because of data access/size
  - Some analyses are computation-heavy and require to be submitted as scripts / run in parallel on server
- If your current workflow with `setwd()` works for you and your colleagues, that's totally fine, but consider future-proofing! ;)

# Recommended and used resources

<https://www.tidyverse.org/blog/2017/12/workflow-vs-script/>

<https://richpauloo.github.io/2018-10-17-How-to-keep-your-R-projects-organized/>

<https://www.rforecology.com/post/organizing-your-r-studio-projects/>

<https://kkulma.github.io/2018-03-18-Prime-Hints-for-Running-a-data-project-in-R/>

<https://rstats.wtf/project-oriented-workflow.html>

<https://appsilon.com/rstudio-shortcuts-and-tips/>

<https://datacornering.com/my-favorite-rstudio-tips-and-tricks/>

<https://happygitwithr.com/>

<https://rstudio.github.io/renv/articles/renv.html>

<https://rstats.wtf/>

Thank you!