

# Krembil Centre for Neuroinformatics

Using big data, artificial intelligence and brain modelling to  
fundamentally change our understanding of mental illness.



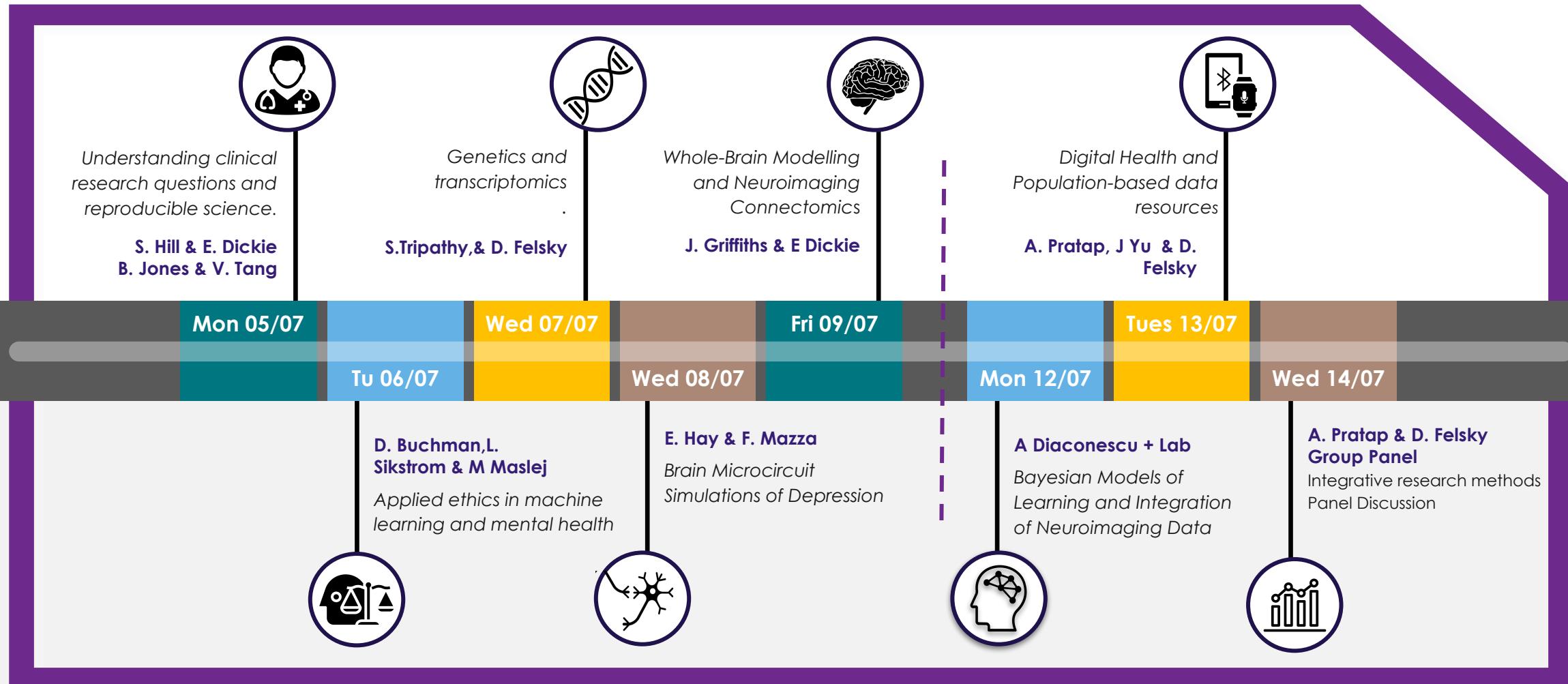
## SUMMER SCHOOL 2020

Day 1

Understanding clinical research questions and reproducible science

Afternoon: Best practices and tools for reproducible science

# Summer School Schedule



# Today's Agenda



**Day 1:**  
Welcome!  
Understanding  
clinical research  
questions and  
reproducible  
science

9:00 am -  
10:30 am

*Lecture 1: Welcome and Orientation + Neuroinformatics Across Scales*  
Erin Dickie + Sean Hill

10:45 am  
- 12:15 pm

*Problems and opportunities in the diagnosis and treatment of major depression*  
Dr Victor Tang & Dr Brett Jones

1:00 pm -  
2:30 pm

*Workshop 1: Guiding principles for FAIR and open science*  
Erin Dickie & Sejal Patel

2:45 pm -  
4:15 pm

*Workshop 2: Tools for Reproducible Science*  
Erin Dickie & Sejal Patel]

# This afternoon

*Guiding  
principles and  
tools for  
reproducible  
science*



## **Erin Dickie Ph.D.**

Lead - Education and Knowledge Transfer  
Krembil Centre for Neuroinformatics, Centre for  
Addiction and Mental Health, Toronto, Ontario  
Twitter: @ErinWDickie Github: @edickie

## **Sejal Patel Ph.D.**

Post-doctoral Fellow - Whole Person Modeling  
Krembil Centre for Neuroinformatics, Centre for  
Addiction and Mental Health, Toronto, Ontario  
Github: @Sejal24

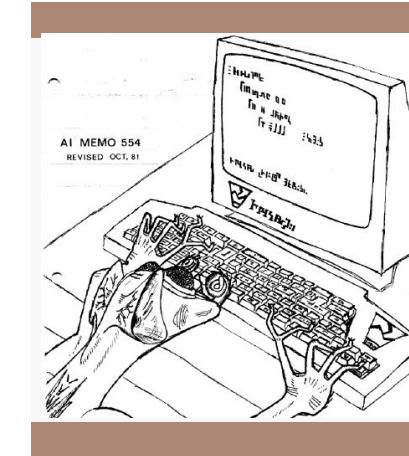
# Teaching Assistants for this section



**Kevin Kadak**  
Whole Brain Modeling  
Lab  
KCNI - CAMH



**Taha Morshedzadeh**  
Whole Brain Modeling  
Lab  
KCNI - CAMH



**Kevin Witczak**  
CAMH Kimel  
TIGRlab  
Github:  
[@kimjetwav](https://github.com/kimjetwav)

# Remember - many ways to engage



(during sessions)  
Use the chat or  
the ask question!



You can always return to the  
session and re-watch the videos  
after the session ends



come chat with us in KCNI  
Summer School Slack :)



virtually meet with us  
in gather.town



Tell us how the session went (post session survey):  
<https://forms.gle/ji18qLMZEZ9L16Ln6>



KCNI School@camh.ca

# Outline for this afternoon

Why reproducible science?

The things you need to know to get through this course:

Versioning and publishing code (github)

Versioning and publishing software (docker)

R with Rmarkdown (walk through)

Python in ipython notebooks (walk through)

-also google colab

The fancy bits:

building your own binder environments

building your own containers (docker & singularity)

# What is reproducible science?

		Data	
		Same	Different
Analysis	Same	Reproducible	Replicable
	Different	Robust	Generalisable

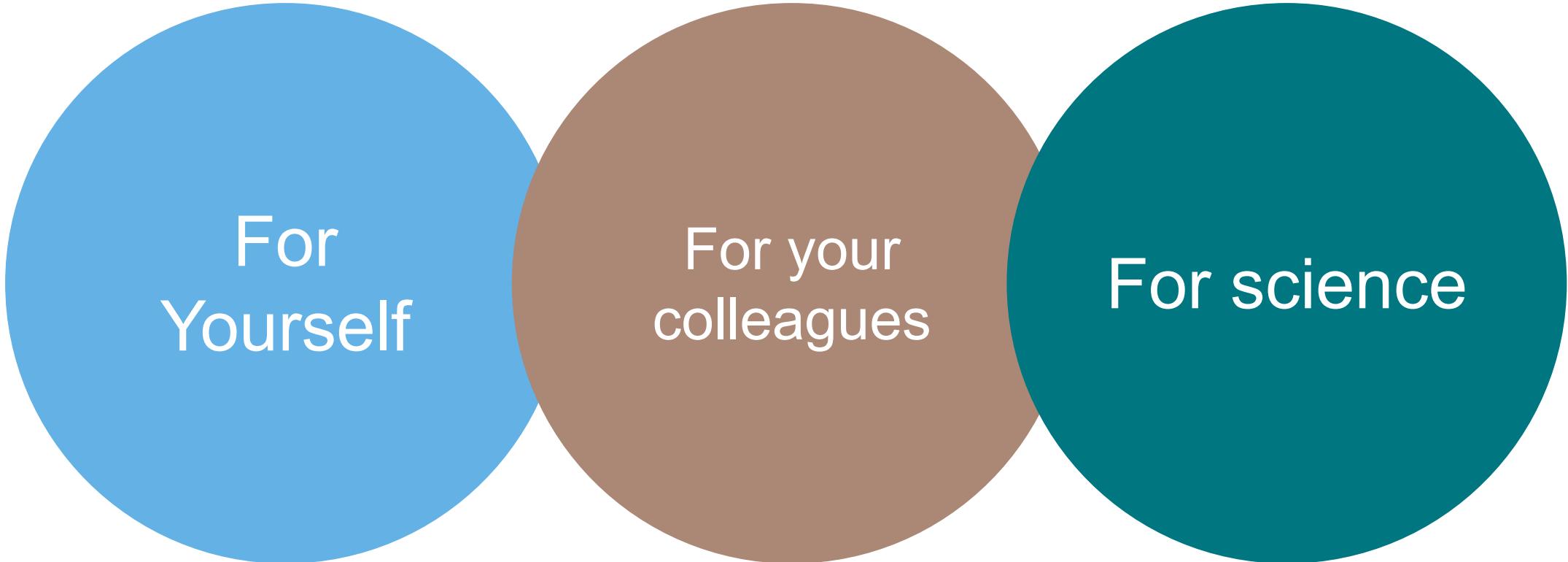
**Fig. 5** How the Turing Way defines reproducible research

# Reproducible Neuroinformatics - What

An article about computational science in a scientific publication is not the scholarship itself, it is merely advertising of the scholarship. The actual scholarship is **the complete software development environment** and the **complete set of instructions which generated the figures**.

Buckheit and Donoho  
WaveLab and Reproducible Research, 1995

# Reproducible Science - Why



For  
Yourself

For your  
colleagues

For science

Your number one  
collaborator is yourself six  
months ago: And they don't  
answer emails

# For Science Friends

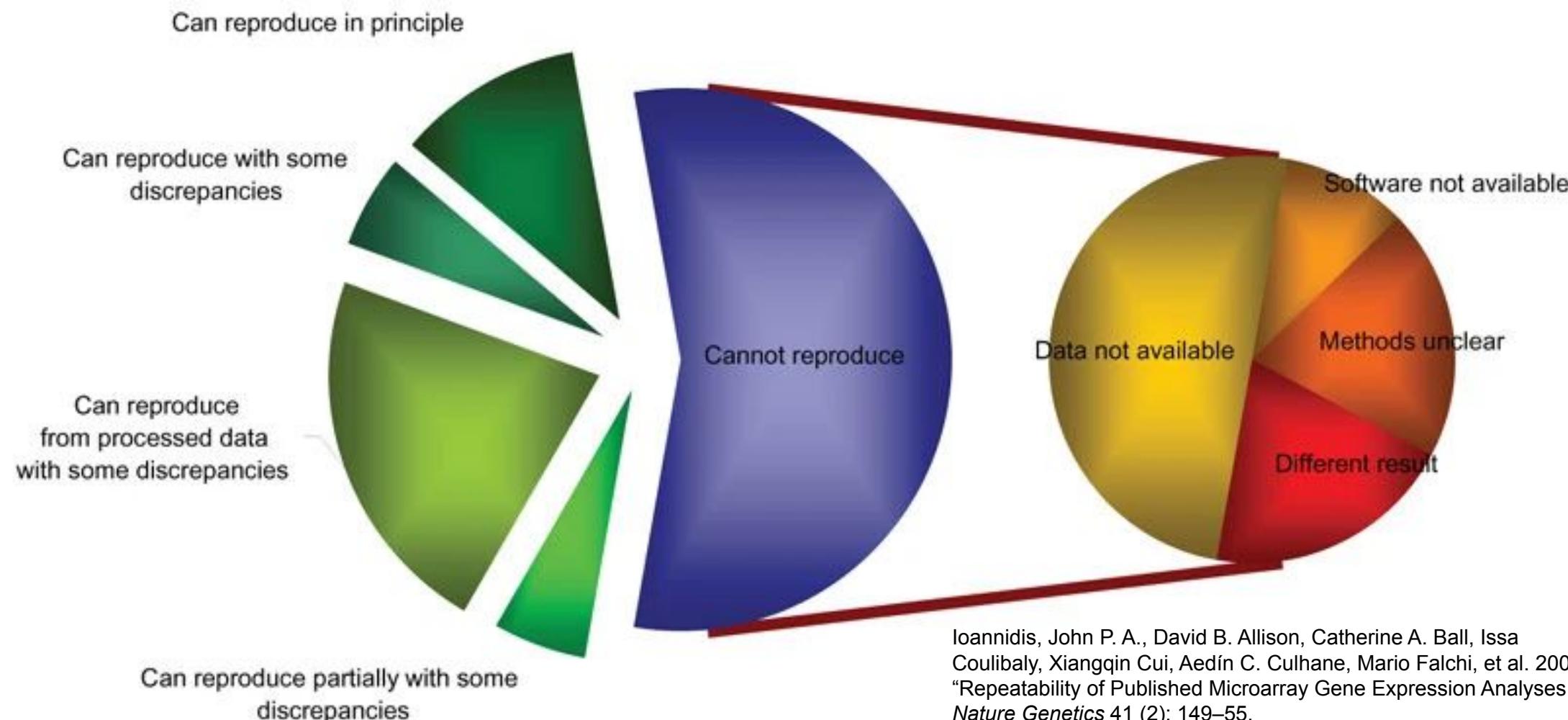
Think of reproducible science as **“tidy” code and data**. By keeping things tidy , labeled and organized you **create a space where you can invite guests**.

Reproducible science leads to **more meaningful collaborations** where collaborators can review your code, learn from it, and **build from you work**.



# Reproducible Science - for science

Tested the reproduction of data analyses in 18 articles on microarray-based gene expression profiling published in *Nature Genetics* in 2005–2006.



# Communicating to your peers

**Problem:** today's scientific papers can fail to communicate all details of the methods needed to reproduce the study.

How to draw an owl

1.



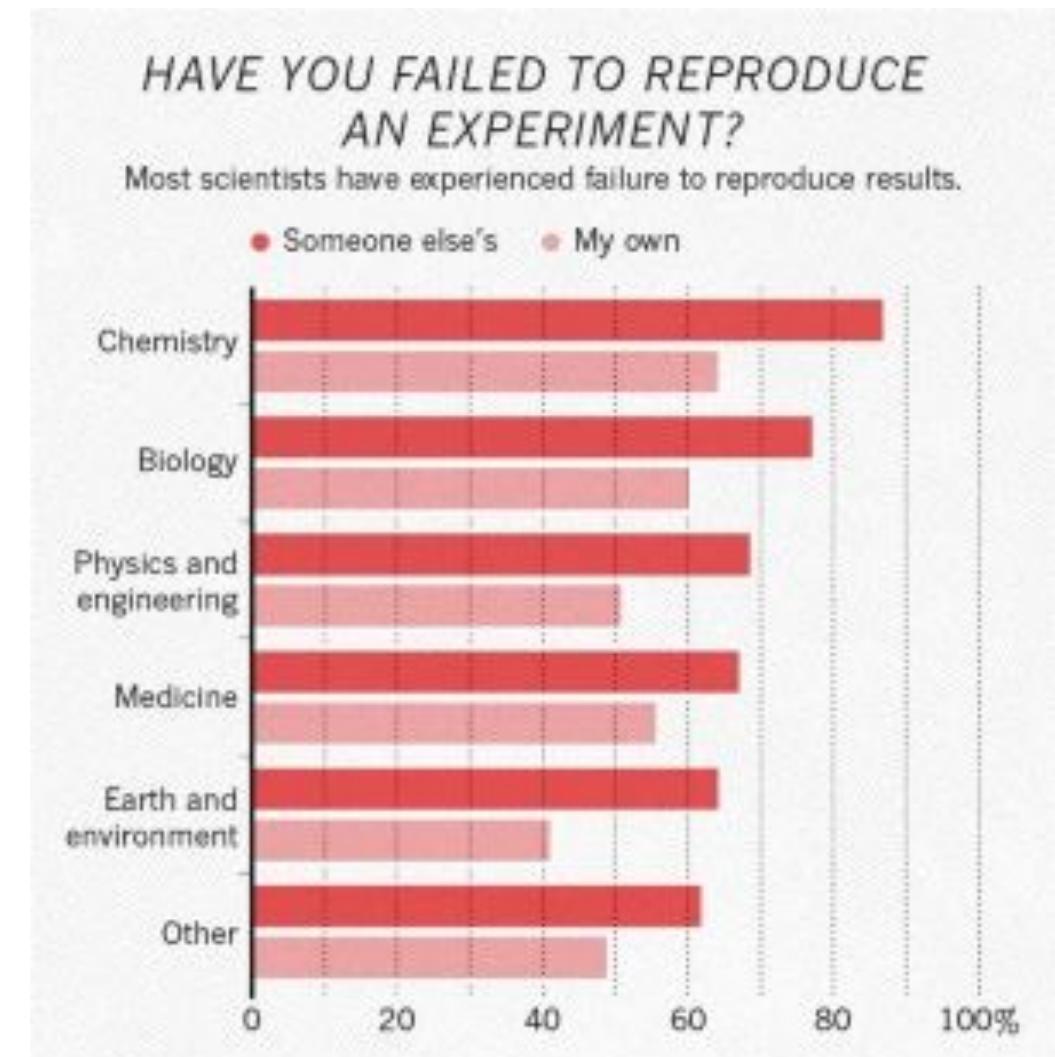
2.



1. Draw some circles

2. Draw the rest of the fooling owl

Baker, Monya. 2016.  
“1,500 Scientists Lift the  
Lid on Reproducibility.”  
*Nature* 533 (7604):  
452–54.



# Problems and Suggested Solutions

Poldrack, Russell A., Chris I. Baker, Joke Durnez, Krzysztof J. Gorgolewski, Paul M. Matthews, Marcus R. Munafò, Thomas E. Nichols, Jean-Baptiste Poline, Edward Vul, and Tal Yarkoni. 2017. "Scanning the Horizon: Towards Transparent and Reproducible Neuroimaging Research." *Nature Reviews Neuroscience* 18 (2): 115–26. <https://www.ncbi.nlm.nih.gov/pmc/articles/PMC6910649/>

Problems	Suggested Solutions
Low statistical power	(one or many) increasing the amount of available data for analysis
Flexibility and exploration in data analysis	pre-registration of methods and analysis plans
Multiple comparisons	sharing of both thresholded and un-thresholded result maps (for meta analysis)
Software errors "As complexity of a software program increases, the likelihood of undiscovered bugs quickly reaches certainty"	avoid the trap of the 'not invented here' philosophy: when the problem at hand can be solved using software tools from a well-established project, these should be chosen instead of re-implementing the same method in custom code
Insufficient study reporting	Because the computer code is often necessary to understand exactly how a data set has been analysed, releasing the analysis code is particularly useful and should be standard practice.

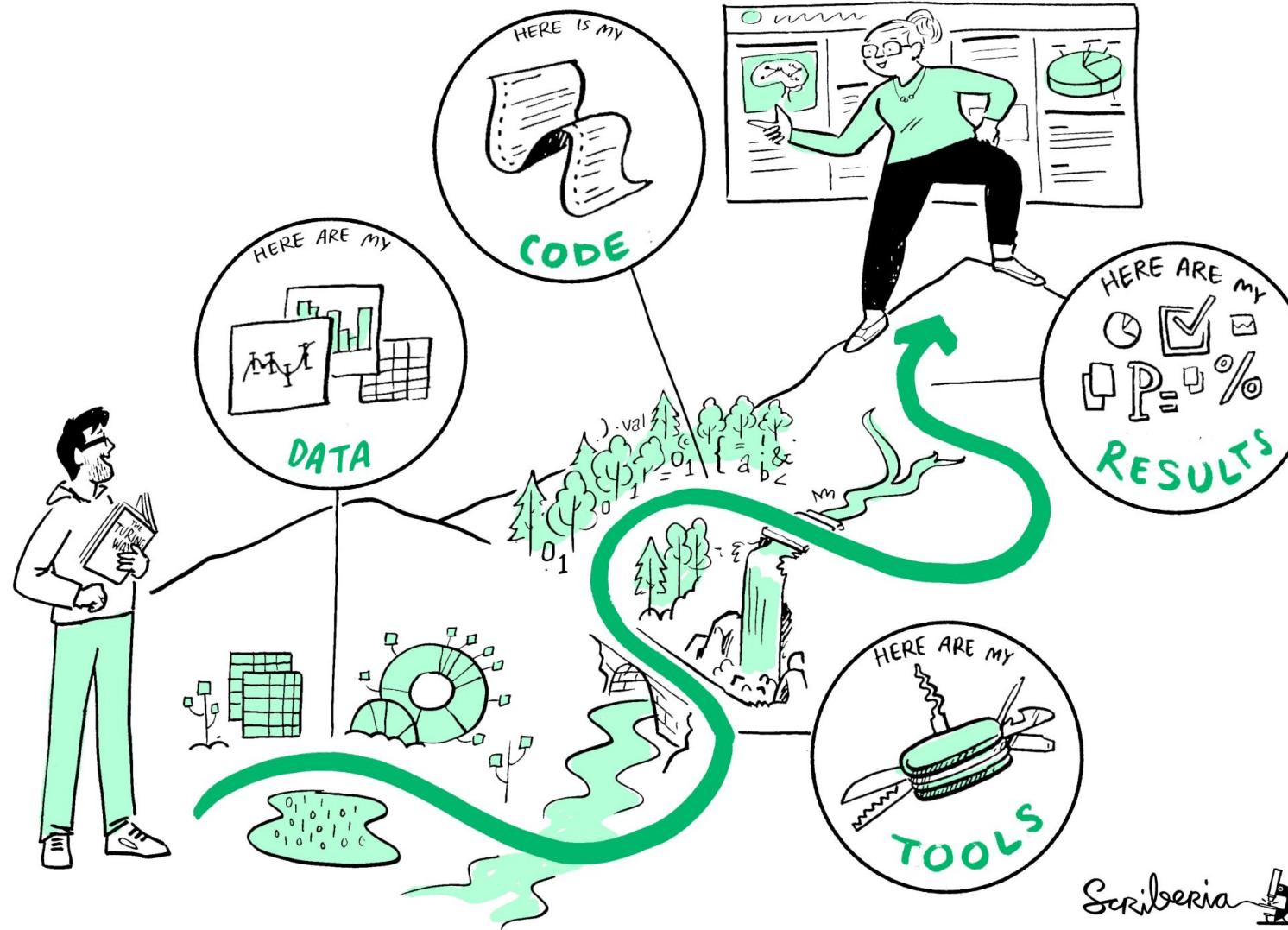
## Towards the [neuroimaging] paper of the future

**“All code for data collection and analysis would be stored in a version-control system and would include software tests to detect common problems”**

“The repository would use a **continuous integration system** to ensure that each revision of the code passes appropriate software tests.”

**“The entire analysis workflow** (including both successful and failed analyses) would be completely automated in a workflow engine and **packaged in a software container** or virtual machine **to ensure computational reproducibility.** “

# The many bits of a project



**Fig. 3** *The Turing Way* project illustration by Scriberia. Used under a CC-BY 4.0 licence.  
DOI:  
[10.5281/zenodo.3332807](https://doi.org/10.5281/zenodo.3332807).

# FAIR data principles



Fig. 31 The Turing Way project illustration by Scriberia. Used under a CC-BY 4.0 licence. DOI: [10.5281/zenodo.3332807](https://doi.org/10.5281/zenodo.3332807).

**Findable:** The first step in (re)using data is to find them! Descriptive metadata (information about the data such as keywords) are essential.

**Accessible:** Once the user finds the data and software they need to know how to access it. Data could be openly available but it is also possible that authentication and authorisation procedures are necessary.

**Interoperable:** Data needs to be integrated with other data and interoperate with applications or workflows.

**Reusable:** Data should be well-described so that they can be used, combined, and extended in different settings.

0. **Organise your project as if you have no memory**
1. **Raw data is read-only. (i.e. never modify your source data)**
2. **Name files for both humans and computers to understand**
3. **Always know where your data comes from**
4. **Track all the changes the humans or computers make. Use version control.**

# Data Naming

Here are some tips for naming files within a research project, which are both human- and machine-readable [Cow20][Hod15]:

- Name your files consistently
- Keep it short but descriptive
- Avoid special characters or spaces to keep it machine-compatible
- Use capitals or underscores to keep it human-readable
- Use consistent date formatting, for example ISO 8601: YYYY-MM-DD to maintain default order
- Include a version number when applicable
- Share/establish a naming convention when working with collaborators
- Record a naming convention in your data management plan

[the turing way - <https://the-turing-way.netlify.app/project-design/filenaming.html>]

# Beware the WOMBAT

W aste  
O f  
M oney  
B rains  
A nd  
T ime

Do not reinvent the wheel

If a (well-maintain) tool or pipeline exists for your analysis - use it

- don't recode your own

If a data naming standard exists for your data type

- use it
- don't invent you own

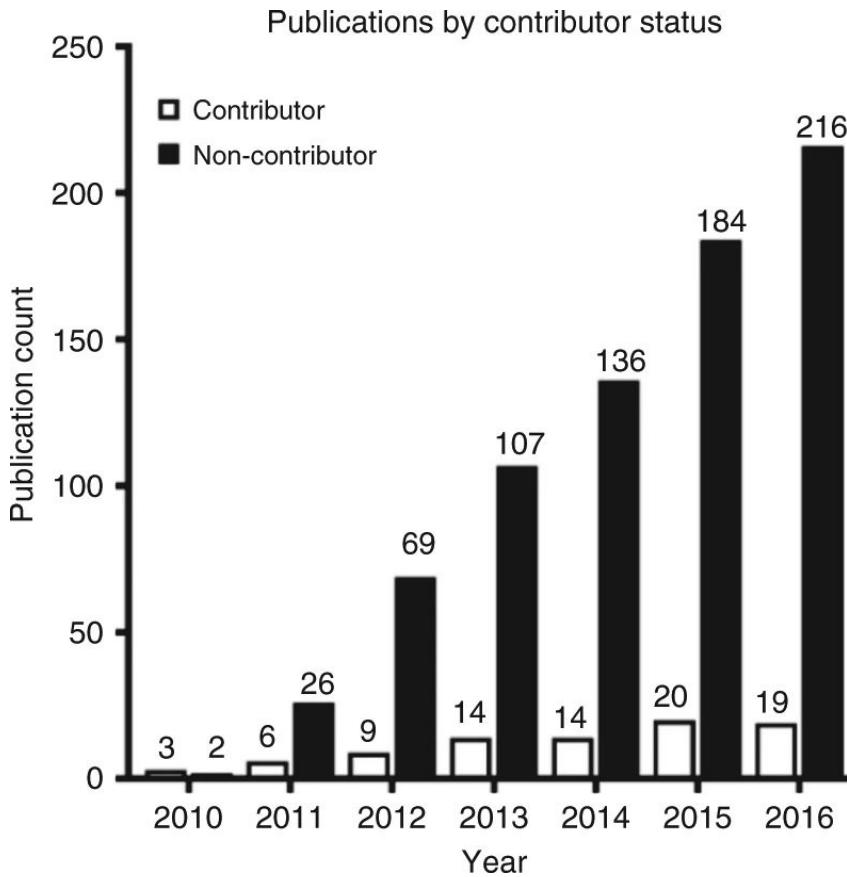
- Michael Hanke @eknahm,  
BrainHack Leipzig 2012

# If it exists - use a data standard!

International Neuroinformatics Coordinating Facility (INCF) hosts a curated list of data standards that are useful our research  
-<https://www.incf.org/resources/sbps>

- Brain Imaging Data Structure (BIDS) for neuroimaging and EEG
- neurodata without borders (NWB) for neurophysiology
- NeuroML for models

# The impact of open data



Database	Cost/ subject	Pheno- typing	Pheno- typing	Clinical	Popu- lation	Difficulty		No. of publica- tions	No. of scans/ subject	\$ Saved	
						Minimal	Compre- hensive	Low	Moderate	High	
FCP	\$1000	x							308	1	101,003,000
ADHD-200	\$2000– 5000			x	x				210	1	526,275,000
NKI-RS	\$3000		x						188	1	70,065,000
ABIDE	\$5000– 10,000				x	x			190	1	995,560,000
CoRR	\$2000	x							17	2	70,065,000

Milham, Michael P., R.et al. 2018. “Assessment of the Impact of Shared Brain Imaging Data on the Scientific Literature.” *Nature Communications* 9 (1): 2818.

# Modelling Best Practices

## Open science is a cultural change



These practices and tools can be difficult to incorporate into a full project with multiple stages of analysis...but they are especially useful in teaching

The materials in this course are examples of reproducible science tools.

# Outline for this afternoon

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**The things you need to know to get through this course:**

**Versioning and publishing code (github)**

Versioning and publishing software (docker)

R with Rmarkdown (walk through)

Python in ipython notebooks (walk through)

-also google colab

The fancy bits:

building your own binder environments

building your own containers (docker & singularity)

# Using version control

**version control: ('git')** is a tool for tracking what changes to a folder (usually a folder filled with code) when and by who..

- Like MS Word's “track changes”
  - ...but for code
  - ..and on steroids..



# Git and Github - What and Why?



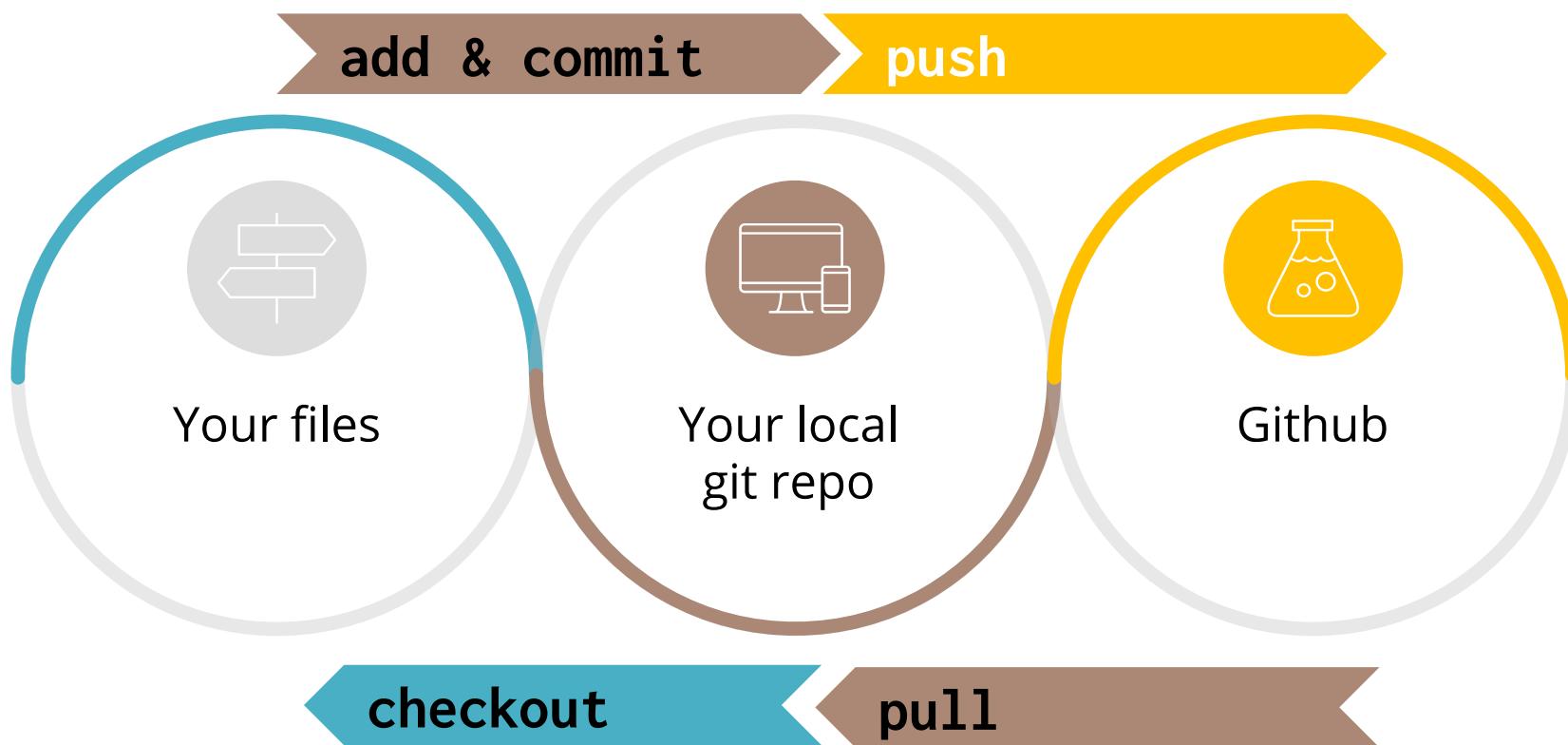
# Github

**Github** is a website where everyone shares their code with themselves, their teams and with the world.

It has a lot of useful features for:

- working with teams
- reading other people's code
- integrating with other platforms
  - continuous integration (CI) to test for bugs
  - Dockerhub
- hosting documentation websites and wiki's
- releasing versions





### **git status**

List what has not been committed (repo)

### **git diff**

tells you want changes haven't been committed (file)

### **git init**

starts a new repo

### **git clone**

copies a repo from GitHub or GitLab to your local computer

# Git for your own project demo



# Git - make a repo

Step 1: build repo on github/gitlab

Step 2:

```
cd ~/code  
git clone http://link/to/git/repo.git
```

First thing in the morning	git pull
Before coffee break	git commit bin/scriptx.py -m "I changed x"
Before going to lunch	git commit bin/scripty.py -m "modifying y"
Done adding function foo	git commit bin/sciprtx.py -m "now does foo"
Before going home	git push

"Git, the  
toothbrush  
of science"



# Git with friends..

Let go look at the lesson's repo on  
github

- when working with larger groups - it's usually better to put in a little more work to make sure that your changes don't clash with other people's changes
- some projects have "CONTRIBUTING.md" files that lay down some best practice steps

- **"fork"** the repo
  - make a copy under your own user space
- create a new **"branch"** that specific to you change
- make your changes inside the **"branch"**
- when you are done make a **"pull request"** back to the **"upstream"** repo

# Git submodules - what? and why?

**What?** - a git “**submodule**” is link from one git repository to another repository

- *think of the [kcni-school-lessosn](#) repo as an [index](#) of lesson repos*
- [on github](#) you see the “submodule” as a [hyperlink](#) to another git repo ***at a specific commit***
- on your computer - a submodule looks like a [git repo inside another git repo.](#)

**Why?** - it can allow you to keep project **more modular**

- meaning you can use it to link to a **common code** base that is shared across multiple projects
- it can **help git run faster** (for very large projects)

**Get the tutorial scripts (with all the submodules) by typing**

```
git clone --recurse-submodules \
https://github.com/krembilneuroinformatics/kcni-school-less...git
```

**Get the updated tutorial scripts (with all the submodules) after  
cloning**

```
cd kcni-school-less...
git pull --recurse-submodules
```

**But what ...**

**you might find, when navigating the kcni-school-lesson that the submodules folders are empty...this is because the submodule needs to he “updated”**

```
cd kcni-school-lessons
cd day1
git submodule init
git submodule update example-python-repo
```

# Git submodules - to create your own

A good tutorial for everything:

<https://git-scm.com/book/en/v2/Git-Tools-Submodules>

*For example here's some code to add SPM12 toolbox code into kcni-school-lesson's day 6 code.*

## **Get add a submodule to your repo**

```
cd kcni-school-lessons/day6/toolboxes  
git submodule add https://github.com/spm/spm12.git
```

## **Get updated the submodule content to the newest commit**

```
cd kcni-school-lessons  
git submodule update --remote spm12
```

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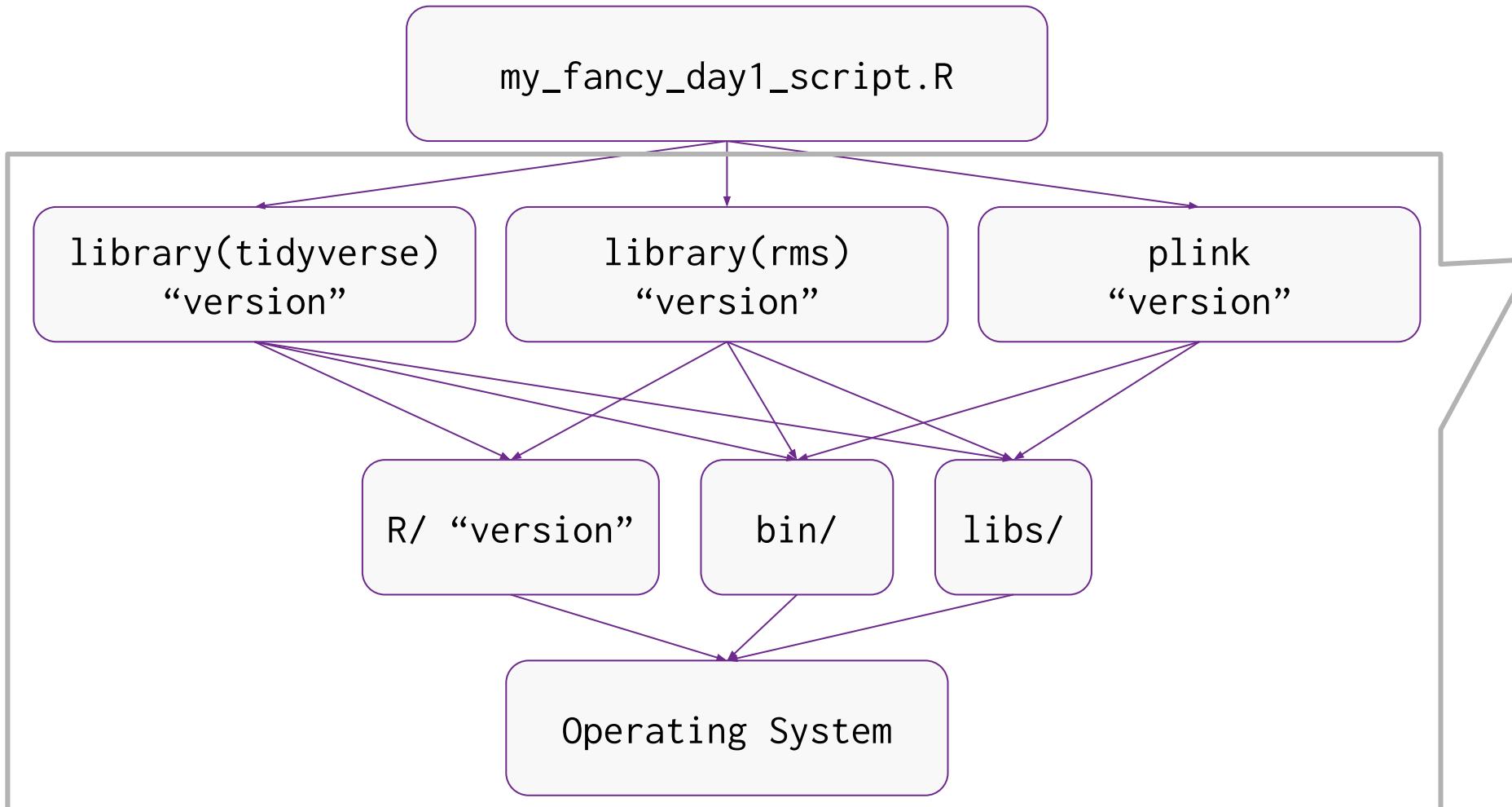
-also google colab

The fancy bits:

building your own binder environments

building your own containers (docker & singularity)

# Containers - Why

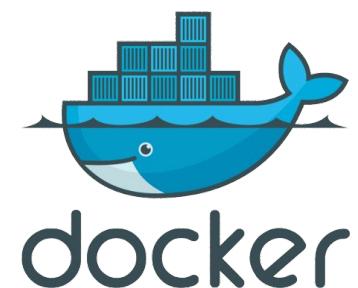
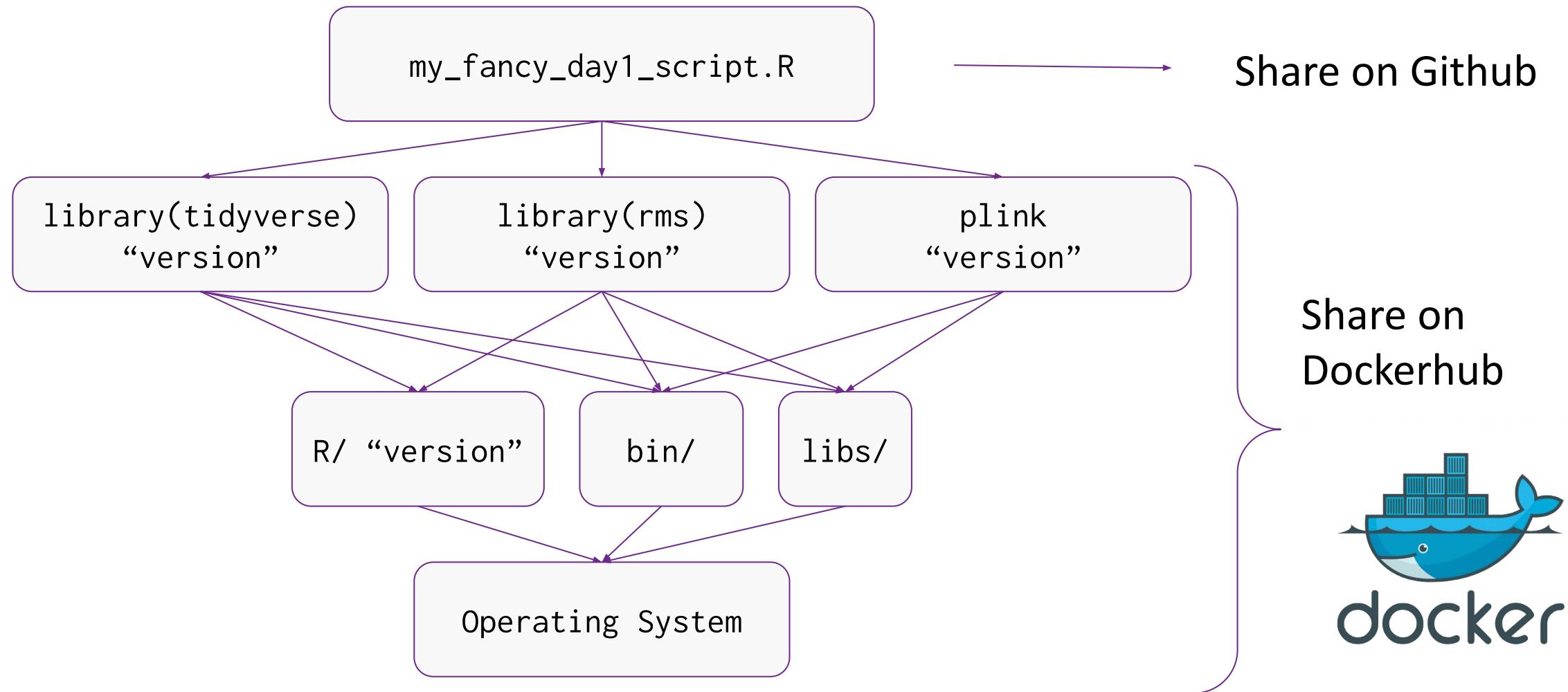


Every box here is something that could be installed differently (*or not at all*) by the next user

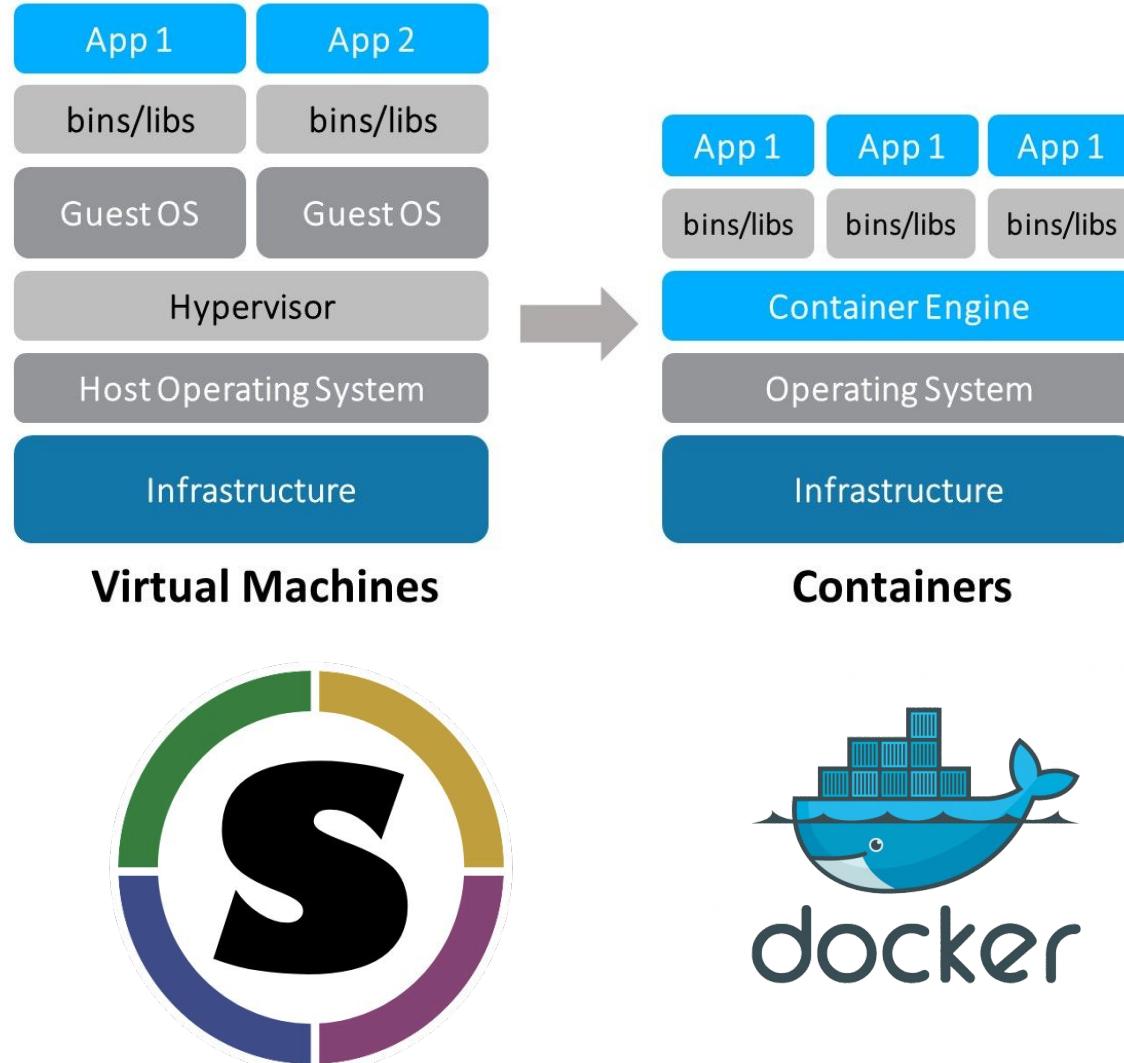
This will cause `my_fancy_day1_script.R` to:

- 1) crash/not work at all
- 2) produce unexpected/different results
- 3) maybe still work?

# Reproducible Neuroinformatics - Solution



# Containers - what and why?



**Docker** - is a tool for sharing software + the dependencies

- the install instructions are stored script called “Dockerfile”
- it’s like a virtual machine
  - without a display
  - that takes up a little less disk space
  - that can be installed in one line

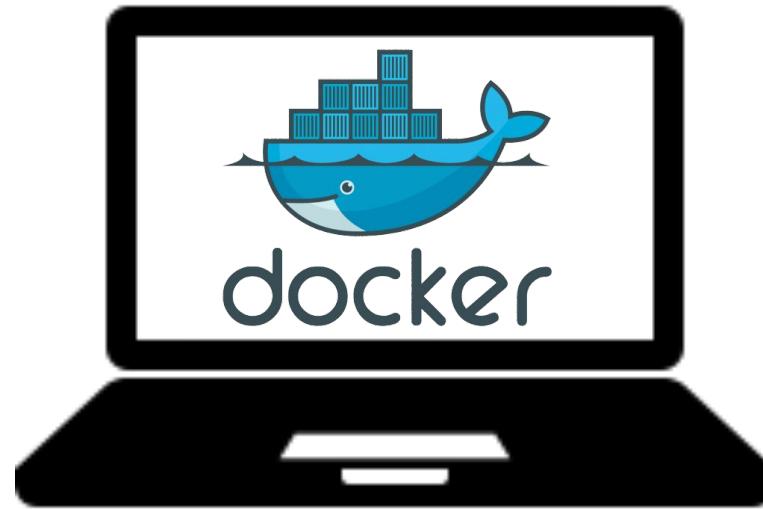
Some Docker vocabulary

- **image**: your install of the software
- **container**: one instance of that software that is *usually* still running.

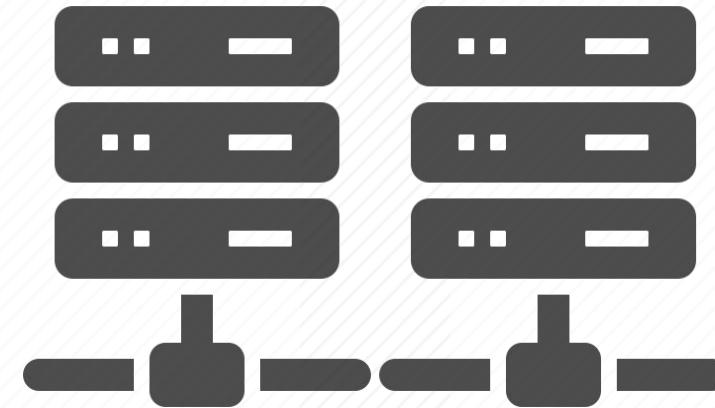
**Dockerhub** is a website that hosts docker images.  
So that anyone - anywhere in the world can run it!

# Containers for scaling up analysis

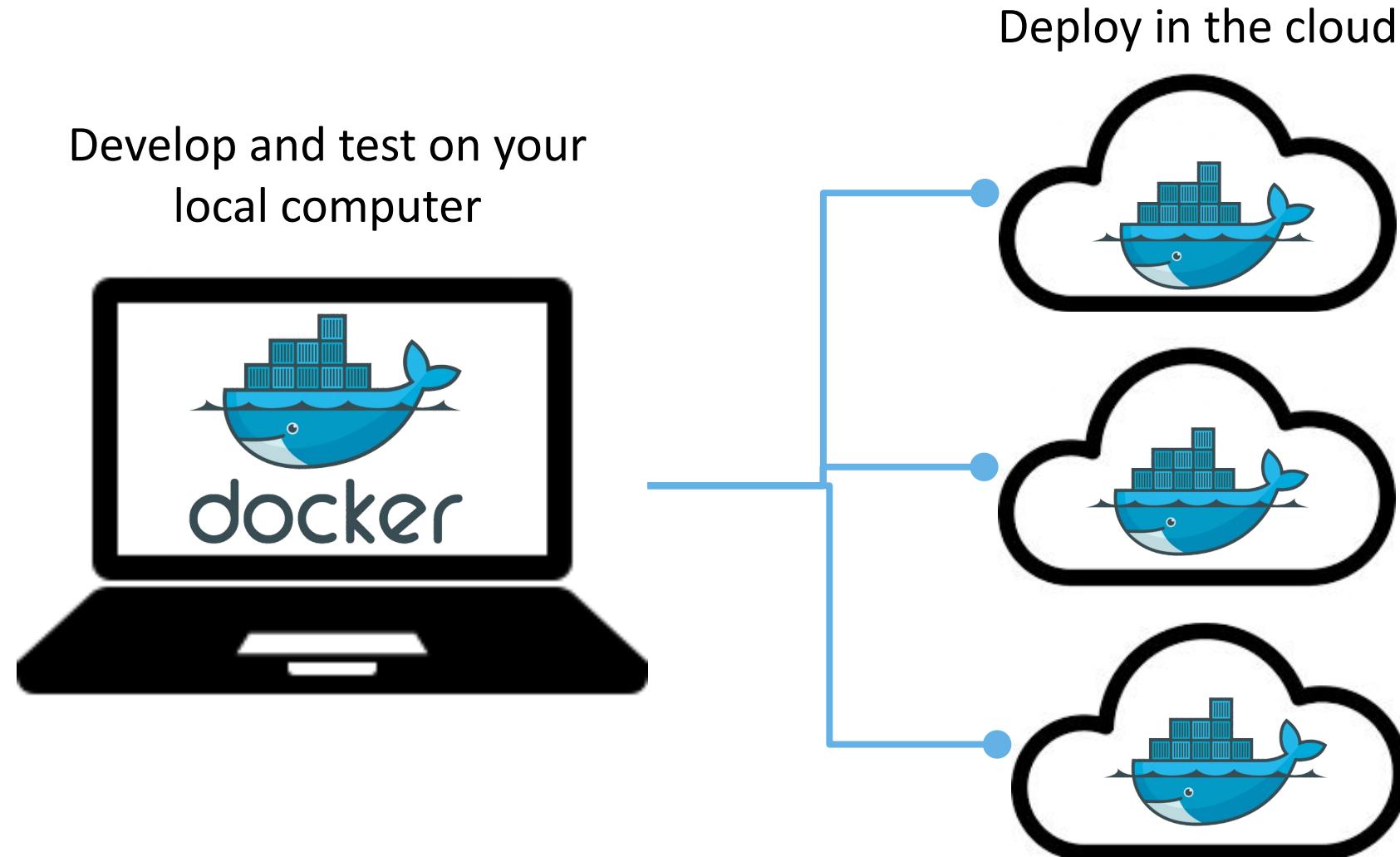
Develop and test on your  
local computer



Deploy software on a high  
performance cluster



# Containers for medical science



There is some speculation that (secure) cloud computing will become important of health research

# How to run our code



## Run on your Local Computer using Docker

- will use less internet bandwidth while you watch the stream
- you will have a copy of the files locally
- requires installation of Docker Desktop



## Run on the web with Binder

- no local installation needed
- may take some time to boot up
- limited resources for the computer



## Run on SciNet using a guest account

- no local installation needed
- a few extra set-up steps needed
- good compute resources

**To interact with the school lesson code on your local computer you will need:**



<https://www.docker.com/products/docker-desktop>

- installs on window, mac or linux

A terminal for pulling/cloning data from github.

-mac os terminal or WSL work

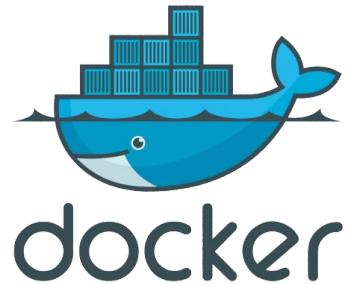


- On windows - gitbash is an option

**Also useful, but not necessary:**



# Step 1: Install Docker Desktop



Installing Docker *should* not be harder than installing any other program on your computer.

Download link and install instructions at:

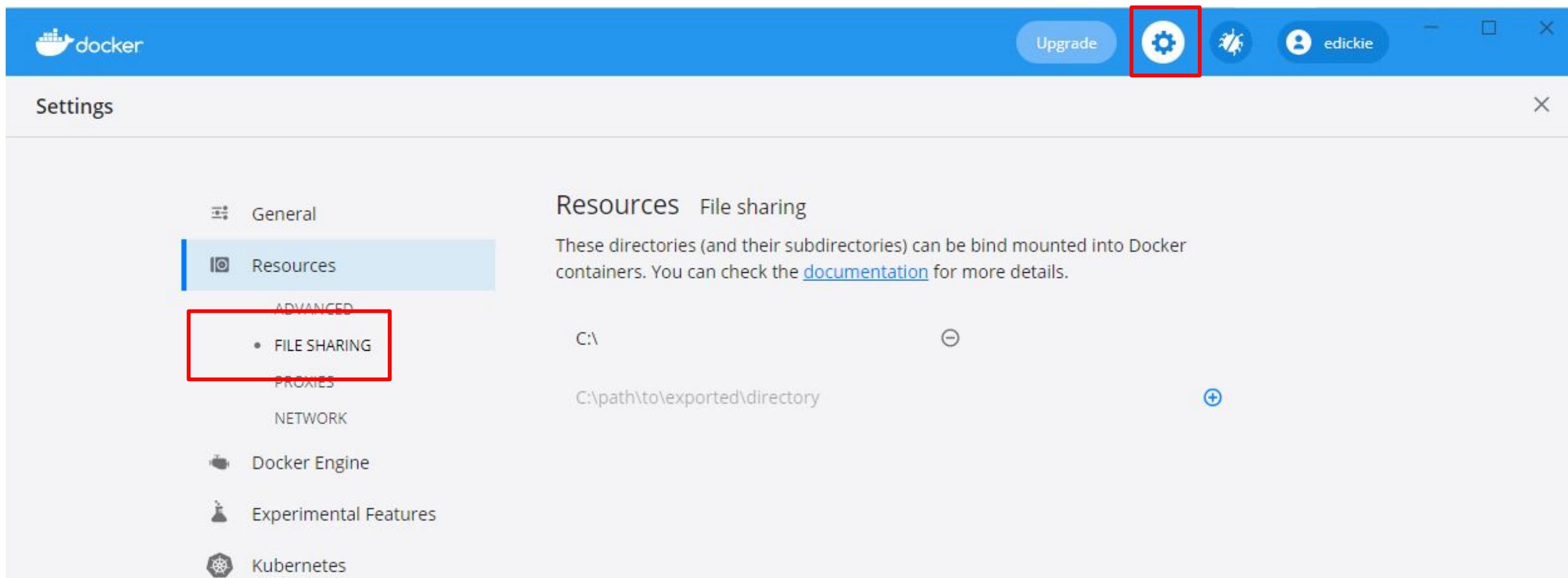
<https://www.docker.com/products/docker-desktop>.

To check your install open up a terminal (in windows this is Powershell or WSL) and type:

```
docker run hello-world
```

# Docker Desktop install gotcha's

1. On Windows - you need to enable Hyper-V or WSL virtualization
2. You also need “Share the drive” with docker.
  - a. Settings->Resources->File Sharing



# If this fails! Fear not - we have a plan!

***If you can't install Docker on your local computer (because you probably don't have enough administrative rights - or you don't have enough space on your home computer) We have a plan!***

***You can run the software on the binder instance or SciNet teach cluster!***

Instructions are available at: <https://github.com/edickie/kcni-school-lessons>

# Step 2: download and run the KCNIschool docker

In the same terminal window where you typed “docker pull hello-world” now type:

```
git clone --recurse-submodules \
  https://github.com/krembilneuroinformatics/kcni-school-lessosn.git
cd kcni-school-lessosn
docker compose up rstudio
```

Then you should see lots of things happening! What is happening? -  
*docker is downloading ~ 5G of software for our lessons into an “image”*

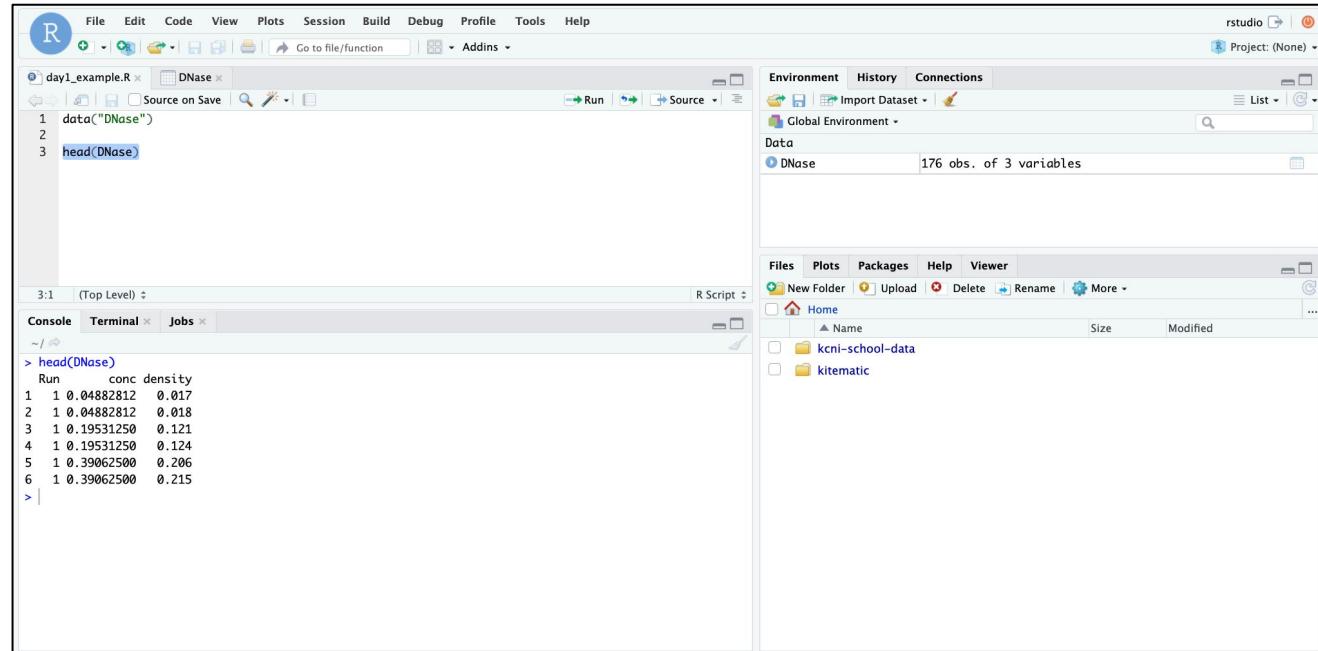
Copy and paste this line from:

<https://github.com/edickie/kcni-school-lessosn/tree/master/envs/README.md>

# Step 3: open rstudio in browser

After typing  
`cd kcni-school-lessos`  
`docker compose up rstudio`

You will finally see the  
message [services.d] done.  
point you browser to:  
<http://localhost:8787/>



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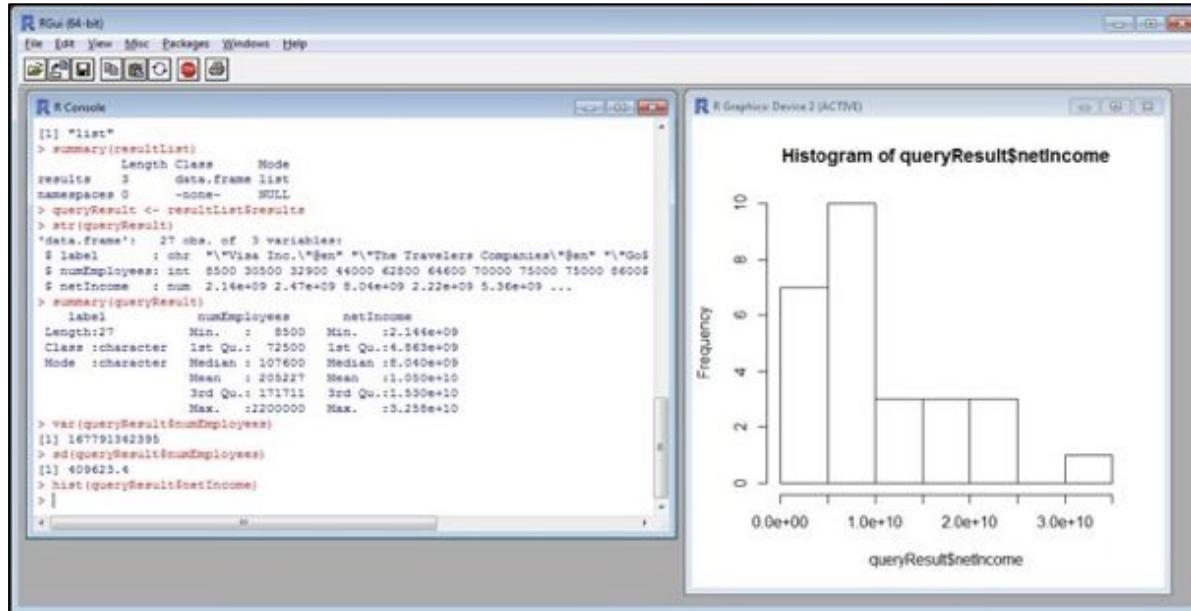
building your own containers (docker & singularity)

**R is a language and environment for statistical computing and graphics for data visualization**

- Similar to S programming language and environment



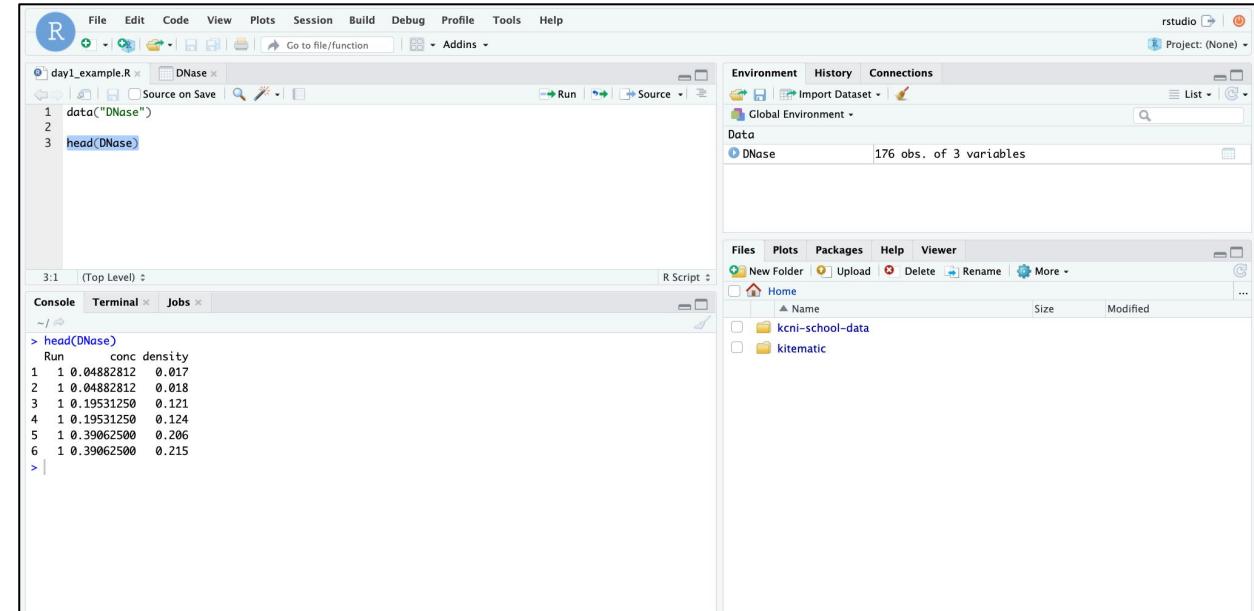
## Old way of coding in R



The screenshot shows the RGui interface. On the left, the R Console window displays R code and its output. The code includes loading a dataset, summarizing it, and creating a histogram. The histogram, titled "Histogram of queryResult\$netIncome", shows the distribution of net income values. The x-axis is labeled "queryResult\$netIncome" and ranges from 0.0e+00 to 3.0e+10. The y-axis is labeled "Frequency" and ranges from 0 to 10. The plot has several bars, with the highest frequency around 1.0e+10.

```
[1] "list"
> summary(resultList)
  Length Class      Node
results  3   data.frame list
namespaces 0   -none-
NULL
> queryResult <- resultList$results
> str(queryResult)
'data.frame': 27 obs. of 3 variables:
$ label : chr "\\"Visa Inc.\\"@en "\\"The Travelers Companies\\"",@en "\\"GoE
$ numEmployees: int 5500 30500 32900 44000 42500 64600 70000 75000 56005
$ netIncome : num 2.14e+09 2.47e+09 8.04e+09 2.22e+09 5.36e+09 ...
> summary(queryResult)
  label      numEmployees      netIncome
Length:27    Min.   : 8500   Min.   :-2.144e+09
Class :character  1st Qu.: 72500  1st Qu.:4.863e+09
Mode  :character  Median : 107600  Median :8.040e+09
               Mean  : 205227  Mean  :1.050e+10
               3rd Qu.: 171731  3rd Qu.:1.550e+10
               Max.  :2200000  Max.  :3.255e+10
> var(queryResult$numEmployees)
[1] 167791342395
> sd(queryResult$numEmployees)
[1] 409623.4
> hist(queryResult$netIncome)
>
#
```

## New way of coding R



The screenshot shows the RStudio interface. On the left, the Script Editor window contains a script named "day1\_example.R" with the following code: `data("DNase")` and `head(DNase)`. The right side of the interface includes the Environment pane (showing the "DNase" dataset), the History pane (empty), and the Connections pane (empty). Below the main panes, the Files pane shows a project structure with folders "knci-school-data" and "kmetic". The bottom section features the Console, Terminal, and Jobs panes, with the Console pane currently active and showing the results of the `head(DNase)` command.

```
File Edit Code View Plots Session Build Debug Profile Tools Help
R Graphics: Device 2 (ACTIVE)
Histogram of queryResult$netIncome
Frequency
0 10 20 30 40 50 60 70 80 90 100
0.0e+00 1.0e+10 2.0e+10 3.0e+10
queryResult$netIncome
R
File Edit Code View Plots Session Build Debug Profile Tools Help
day1_example.R DNase
Source on Save Run Source
1 data("DNase")
2
3 head(DNase)

3:1 (Top Level) R Script
Console Terminal Jobs
> head(DNase)
Run conc density
1 1 0.04882812 0.017
2 1 0.04882812 0.018
3 1 0.19531250 0.121
4 1 0.19531250 0.124
5 1 0.39062500 0.206
6 1 0.39062500 0.215
>
```

# Rstudio Interface and Demo

The image shows the RStudio interface with several components highlighted:

- R Script or code editor:** The left pane displays an R script named "day1\_example.R". A red box highlights the "Run" button in the toolbar above the code area.
- R Console:** The bottom-left pane shows the R console output for a command that prints a data frame. A red box highlights the "Console" tab in the bottom navigation bar.
- R Environment:** The top-right pane shows the global environment with objects like "case\_df1", "case\_df1b", and "case\_df2". A red box highlights the "Plots" tab in the bottom navigation bar.
- File and Graphical Output:** The bottom-right pane displays a grouped bar chart titled "File and Graphical Output". The x-axis is "alcgp" (Alcohol Consumption Group) and the y-axis is "ncases" (Number of cases). The legend indicates age groups: 25-34 (light blue), 35-44 (medium blue), 45-54 (dark blue), 55-64 (light green), 65-74 (medium green), and 75+ (yellow).

Code in the R Script editor:

```
library(tidyverse)
library(ggplot2)

# Summarize data
summarise(case_df1a, ncases, alcgp, agegp)
case_df1b <- case_df1a[order(case_df1a$ncases, decreasing = TRUE),]
head(case_df1b)

# Tidyverse: which age group has the highest number of alcohol consumption +
```

Console output:

```
1 65-74 40-79 17
2 55-64 40-79 9
3 55-64 80-119 9
4 55-64 80-119 8
5 45-54 40-79 6
6 45-54 80-119 6
```

Plot Data (Estimated from chart):

alcgp	25-34	35-44	45-54	55-64	65-74	75+
0-39g/day	1	1	1	4	4	2
40-79	3	3	3	4	4	2
80-119	3	3	3	4	4	1
120+	2	2	2	3	3	2

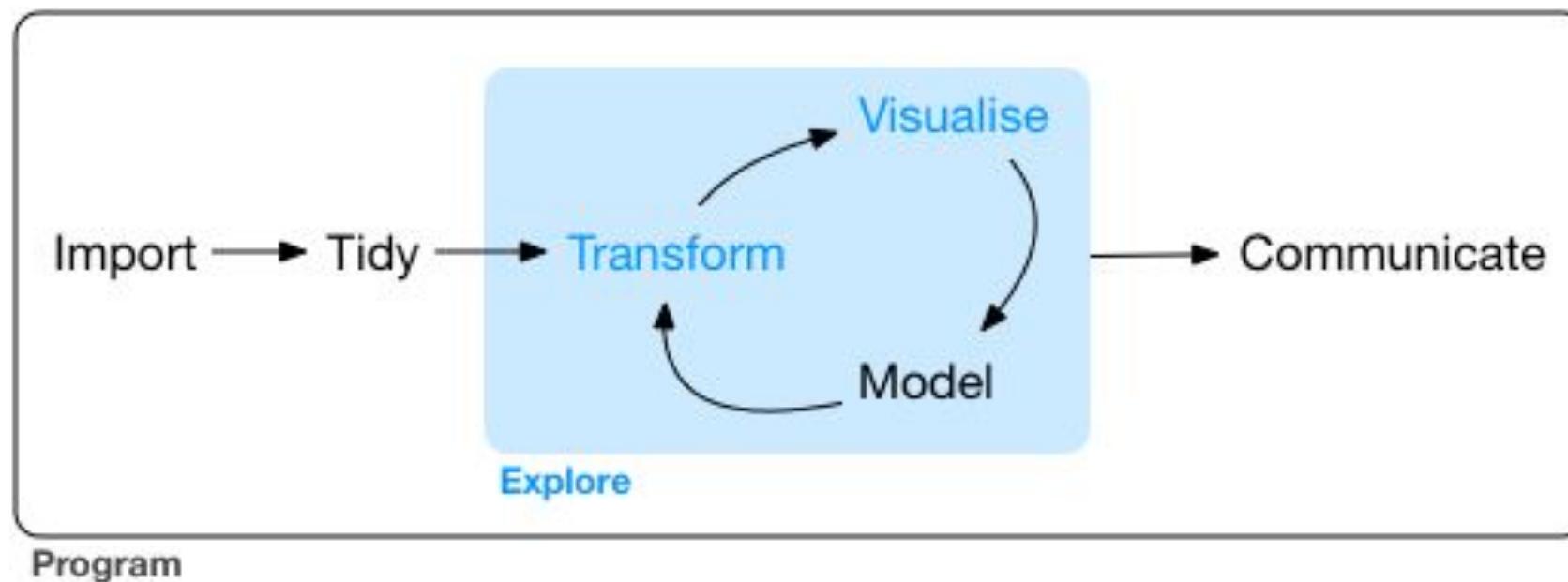
# Tidyverse (dplyr package) vs Base R

- Here are few examples between base R function and the equivalent functions in the dplyr package found within tidyverse
- Note: through the different workshops you will encounter the use of both tidyverse and base R coding style

dplyr	base
arrange(df, x)	df[order(x), , drop = FALSE]
distinct(df, x)	df[!duplicated(x), , drop = FALSE], unique()
filter(df, x)	df[which(x), , drop = FALSE], subset()
mutate(df, z = x + y)	df\$z <- df\$x + df\$y, transform()
pull(df, 1)	df[[1]]
pull(df, x)	df\$x
rename(df, y = x)	names(df)[names(df) == "x"] <- "y"
relocate(df, y)	df[union("y", names(df))]
select(df, x, y)	df[c("x", "y")], subset()
select(df, starts_with("x"))	df[grep(names(df), "x")]
summarise(df, mean(x))	mean(df\$x), tapply(), aggregate(), by()
slice(df, c(1, 2, 5))	df[c(1, 2, 5), , drop = FALSE]

## Long debate in the R community of which way of R coding is better

- Generally Tidyverse is easier to learn than Base R
- Tidyverse follows some logical flow when coding which is easier to understand



## Two main ways to write and execute code

- R script file can be used to write your code and the file extension is .R
  - code and output are on two different panels
- However using R Notebooks is becoming used more often
  - The output of the code is below each code chunk
  - Documenting the code and reporting can be done beside the code using text elements to get a fully formatted
  - Can use other programming language such as Bash or Python
  - Easy to share with collaborator
  - Work well with version control system
  - R Notebook is way to work with R Markdown files

## R Markdown has a file extension .Rmd

- Provides an authoring framework for data science
- Fully formatted document into PDF, HTML or Word
- Combination of:
  - Written in plain text
  - Special characters for text formatting
  - R code within it to produce outputs such as table and plots
- Generate high quality reports that can be shared with an audience

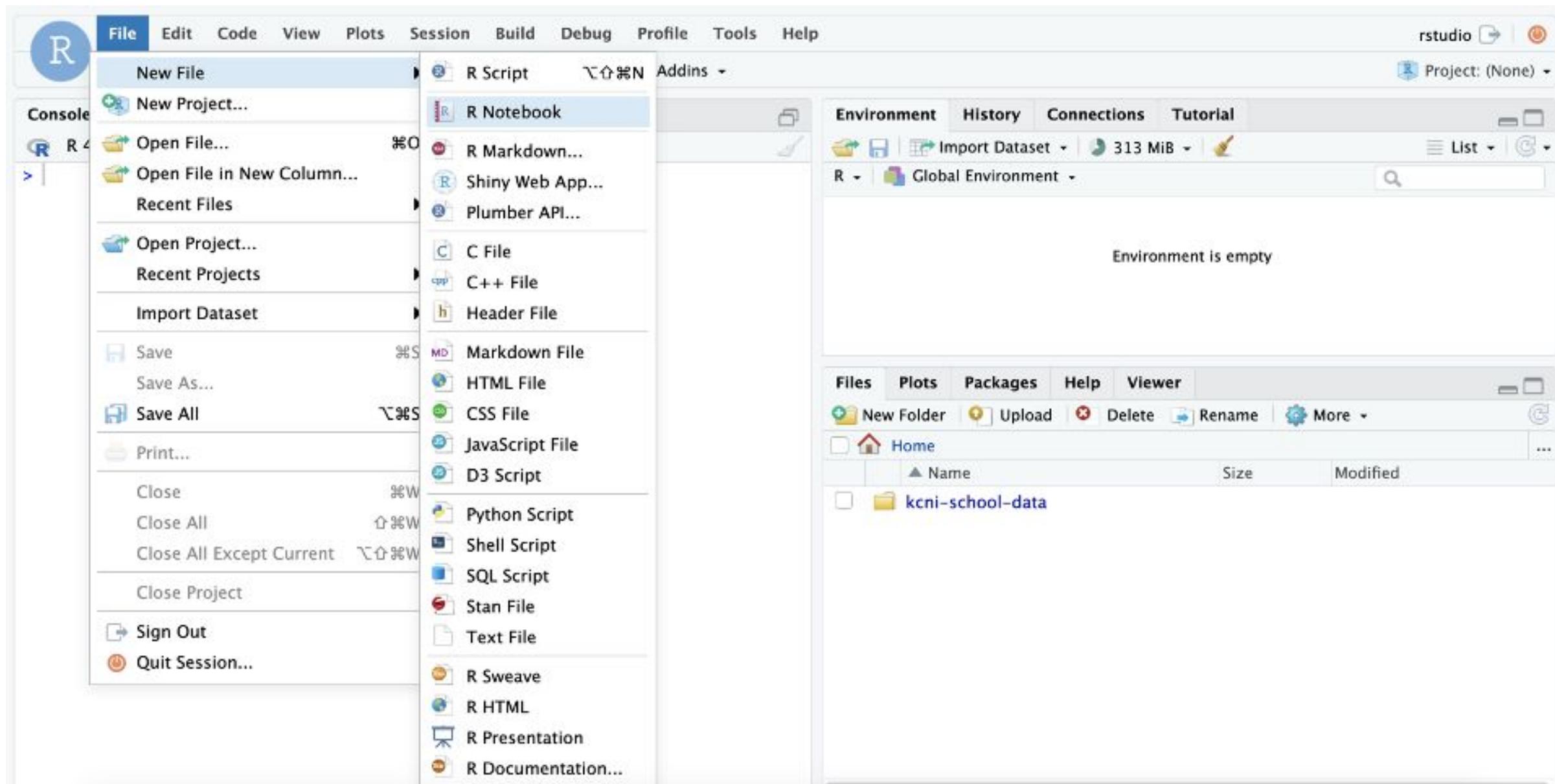


# R Script Demo

The screenshot displays the RStudio interface with the following components:

- File Menu:** File, Edit, Code, View, Plots, Session, Build, Debug, Profile, Tools, Help.
- Project Bar:** Shows "rstudio" and "Project: (None)".
- Script Editor:** Displays two files: "Day1\_script\_example.R" and "Day1\_Demo.Rmd". The "Day1\_script\_example.R" file contains R code for loading libraries and a dataset, followed by a summary of the "esoph" dataset. The "Day1\_Demo.Rmd" file is currently selected.
- Console:** Shows the R command-line interface output, including the loading of tidyverse and ggplot2 packages, the selection of the esoph dataset, and its summary statistics.
- Environment:** Shows the global environment with four data frames: case\_df1a, case\_df1b, case\_df2, and esoph, each with 88 observations and varying numbers of variables.
- Files:** A file browser showing the directory structure: Home > kcni-school-data > kcni-school-lessos > day1 > example-r-repo. It lists files like .github, Day1\_Demo.html, Day1\_Demo.Rmd, Day1\_script\_example.R, index.ipynb, install.R, LICENSE, README.md, and runtime.txt, all modified on Jul 5, 2021, at 9:57 AM.

# R Notebook Example + Demo



# R Notebook Example + Demo

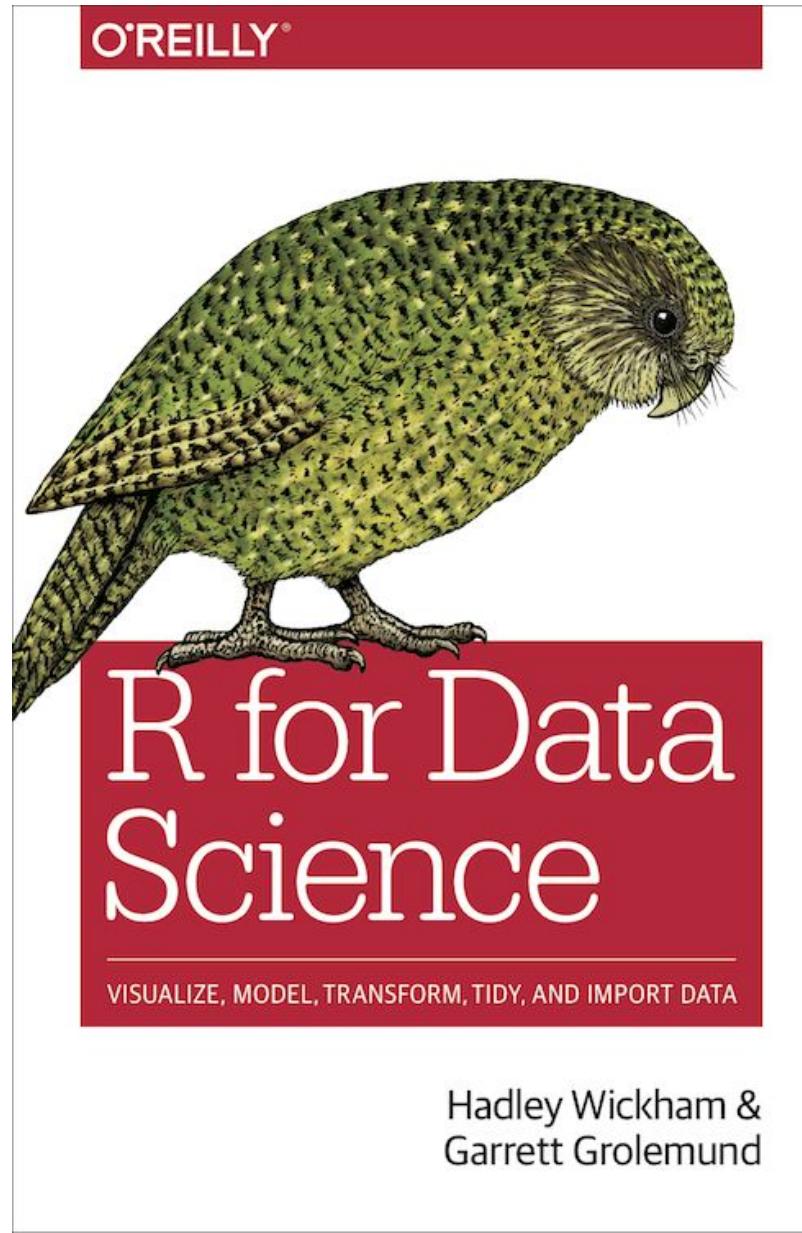
The screenshot shows the RStudio interface with the following details:

- Top Bar:** File, Edit, Code, View, Plots, Session, Build, Debug, Profile, Tools, Help.
- Project Bar:** Project: (None).
- Left Sidebar:** Shows an "Untitled1" file open. The code content is as follows:

```
1 ---  
2 title: "R Notebook"  
3 output: html_notebook  
4 ---  
5  
6 This is an [R Markdown](http://rmarkdown.rstudio.com) Notebook. When you execute code within the notebook, the results appear beneath the code.  
7  
8 Try executing this chunk by clicking the *Run* button within the chunk or by placing your cursor inside it and pressing *Ctrl+Shift+Enter*.  
9
```

- Right Sidebar:** Environment pane shows "Environment is empty".
- Bottom Right:** Buttons for "File" and "New Folder".

# R for Data Science



<https://r4ds.had.co.nz/index.html>

# R Cheatsheets!

The screenshot shows the RStudio interface with the Help menu open. The "Cheatsheets" option is highlighted in blue, indicating it is selected. The "Cheatsheets" section of the help menu is displayed, listing various cheat sheets and quick references.

**Help Menu Options:**

- R Help
- About RStudio
- RStudio Docs
- RStudio Community Forum
- Cheatsheets** (selected)
- Keyboard Shortcuts Help ⌘`
- Markdown Quick Reference
- Roxygen Quick Reference
- Diagnostics

**Cheatsheets List:**

- RStudio IDE Cheat Sheet
- Data Transformation with dplyr
- Data Visualization with ggplot2
- List manipulation with purrr
- Package Development with devtools
- Web Applications with shiny
- Interfacing Spark with sparklyr
- R Markdown Cheat Sheet
- R Markdown Reference Guide
- Browse Cheatsheets...

**Code Editor Content:**

```
library(tidyverse)
library(ggplot2)

data("esoph")
summary(esoph)

# baseR: which age ago has the highest number of alcholo consumption
case_df1a <- esoph[c("agegp", "alcp", "ncases")]
case_df1b <- case_df1a[order(case_df1$ncases, decreasing = TRUE),]
head(case_df1b)

# Tidyverse: which age ago has the highest number of alcholo consumption that are cases
case_df2 <- esoph %>% select(c(agegp,alcp,ncases)) %>% arrange(desc(ncases))
head(case_df2)

#Tidyverse plot the number of cases in each alcogp group by ages
ggplot(case_df2,aes(x=alcp, y=ncases, fill=agegp)) +
```

# Outline for this afternoon

Why reproducible science?

The things you need to know to get through this course:

Versioning and publishing code (github)

Versioning and publishing software (docker)

R with Rmarkdown (walk through)

**Python in ipython notebooks (walk through)**

-also google colab

The fancy bits:

building your own binder environments

building your own containers (docker & singularity)

Jupyter notebooks - let's open the  
one in our example-python-repo

```
docker compose up jupyter
```



# Jupyter file browser interface

The screenshot shows the Jupyter Notebook file browser interface. At the top, there are three tabs: 'home/neuro/kcni-school-data/' (closed), 'step1 - Jupyter Notebook' (active), and 'step2 - Jupyter Notebook'. Below the tabs is a toolbar with icons for back, forward, search, and various file operations. The title bar displays the URL 'http://127.0.0.1:8888/tree/home/neuro/kcni-school-data/day1'. On the left, there is a logo for 'jupyter' and two buttons: 'Quit' and 'Logout'. Below the title bar, there are three tabs: 'Files' (selected), 'Running', and 'Clusters'. A message 'Select items to perform actions on them.' is displayed above the file tree. The 'File Tree' section shows a list of files and folders under the path '/ home / neuro / kcni-school-data / day1'. The files listed are: 'step1.ipynb' (Running, 72 B, a minute ago), 'step2.ipynb' (Running, 555 B, seconds ago), and 'README.md' (0 B, 12 days ago). There is also a folder '...' (seconds ago). The table has columns for Name, Last Modified, and File size. A green arrow points to 'step2.ipynb' with the annotation '← Running Notebook in green'.

	Name	Last Modified	File size
<input type="checkbox"/>	...	seconds ago	
<input type="checkbox"/>	step1.ipynb	a minute ago	72 B
<input type="checkbox"/>	step2.ipynb	Running	555 B
<input type="checkbox"/>	README.md	12 days ago	0 B

# Jupyter file browser interface

The screenshot shows a Jupyter Notebook interface with the following elements:

- Toolbar:** Includes icons for back, forward, search, and various notebook operations. The "Run" button (a play icon) and the "add new cell" button (a plus sign) are highlighted with red boxes.
- Header:** Shows three open notebooks: "home/neuro/kcni-school-data", "step1 - Jupyter Notebook", and "step2 - Jupyter Notebook". The URL is http://127.0.0.1:8888/notebooks/home/neuro/kcni-school-d... .
- Menu Bar:** File, Edit, View, Insert, Cell, Kernel, Widgets, Help.
- Cell Type:** Trusted Python 3.
- Tool Buttons:** Includes icons for file operations like save, copy, paste, and cell controls like run, execute, and kernel restart.
- Content Area:** A markdown cell containing the text "This cell is written in markdown" with a note "(double click on it to edit)".
- Code Cell:** An In [12] cell containing Python code: 

```
python In [12]: ## this cell is python code - the output shows up below
print("Hello World")
```

 The output "Hello World" is displayed below the code.
- Footer:** A section titled "Things we should learn" with a bullet point: "bit of the interface".

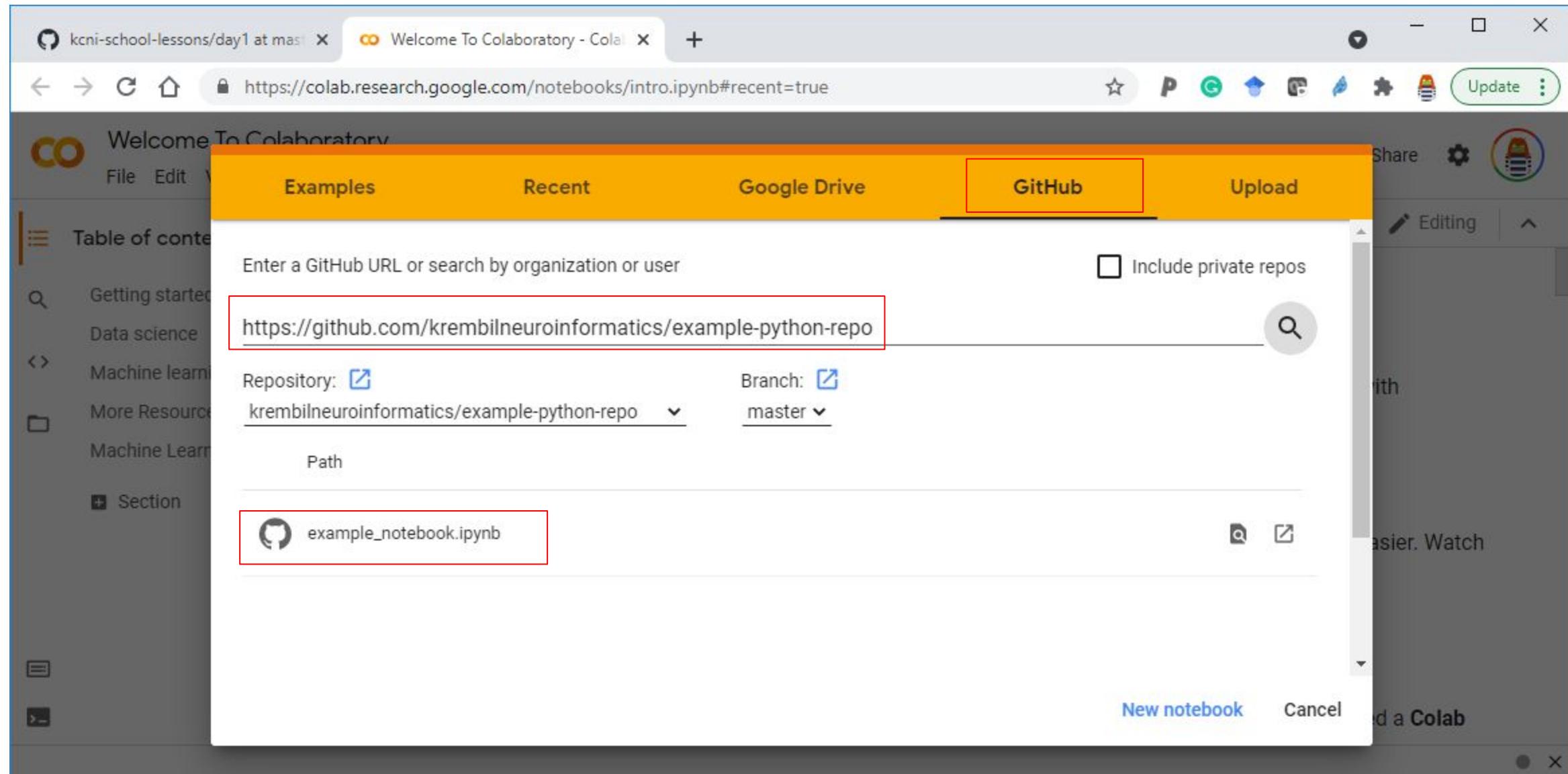
Now let's check out the same  
notebook on google colaboratory  
(colab)

Instructions at:

<https://github.com/krembilneuroinformatics/kcni-school-less...>



# Recap: opening the notebook



# The colab interface (README.md)

The screenshot shows a Google Colab notebook titled "example\_notebook.ipynb". The interface includes a toolbar with file operations like "File", "Edit", "View", "Insert", "Runtime", "Tools", and "Help". Below the toolbar, there are buttons for "+ Code" and "+ Text", and a "Copy to Drive" option. A red box highlights the "+ Code" button with the label "add new cell". The main workspace contains a markdown cell with the text "This cell is written in markdown" and a note "(double click on it to edit)". A red arrow points from this text to the "Run current cell" button. Below this is a Python cell containing the code "# this cell is python code - the output shows up below" and "print('Hello World')". A red box highlights the play button icon in the cell header, and another red arrow points from this icon to the same "Run current cell" button. A red arrow also points from the "Run current cell" button to the Python code in the cell. At the bottom of the workspace, a note says "Note: this notebook uses the seaborn plotting package - which is listed in requirements.txt". The sidebar on the left lists sections: "Things we should learn" (which is collapsed), "bit of the interface", "python block", "markdown blocks", and "plotting". A red arrow points from the "you can collapse (hide) cells" text to the collapsed "Things we should learn" section.

+ Code + Text

add new cell

CO example\_notebook.ipynb

File Edit View Insert Runtime Tools Help

Share

+ Text

Copy to Drive

Connect Editing

This cell is written in markdown

(double click on it to edit)

Run current cell

## this cell is python code - the output shows up below

```
print("Hello World")
```

Note: this notebook uses the seaborn plotting package - which is listed in requirements.txt

Things we should learn

- bit of the interface
- python block
- markdown blocks
- plotting

you can collapse (hide) cells

# colab strengths and caveates

## strengths

- more power, more diskspace and more RAM than binder
- easy to use
- can be left running for days (no timeouts, unlike binder)
- can connect to google drive for more cloud storage space.

## caveats

### to remember for this course

- each colab notebook is an island
  - it is not aware of the other data scripts or notebooks in same folder
- You always start with the same (cleanish) linux/python-3.7.1 environment
  - so you need to install all other software inside your notebook
  - the current env is old so some newer packages may not work

# Outline for this afternoon

Why reproducible science?

The things you need to know to get through this course:

Versioning and publishing code (github)

Versioning and publishing software (docker)

R with Rmarkdown (walk through)

Python in ipython notebooks (walk through)

-also google colab

**The fancy bits:**

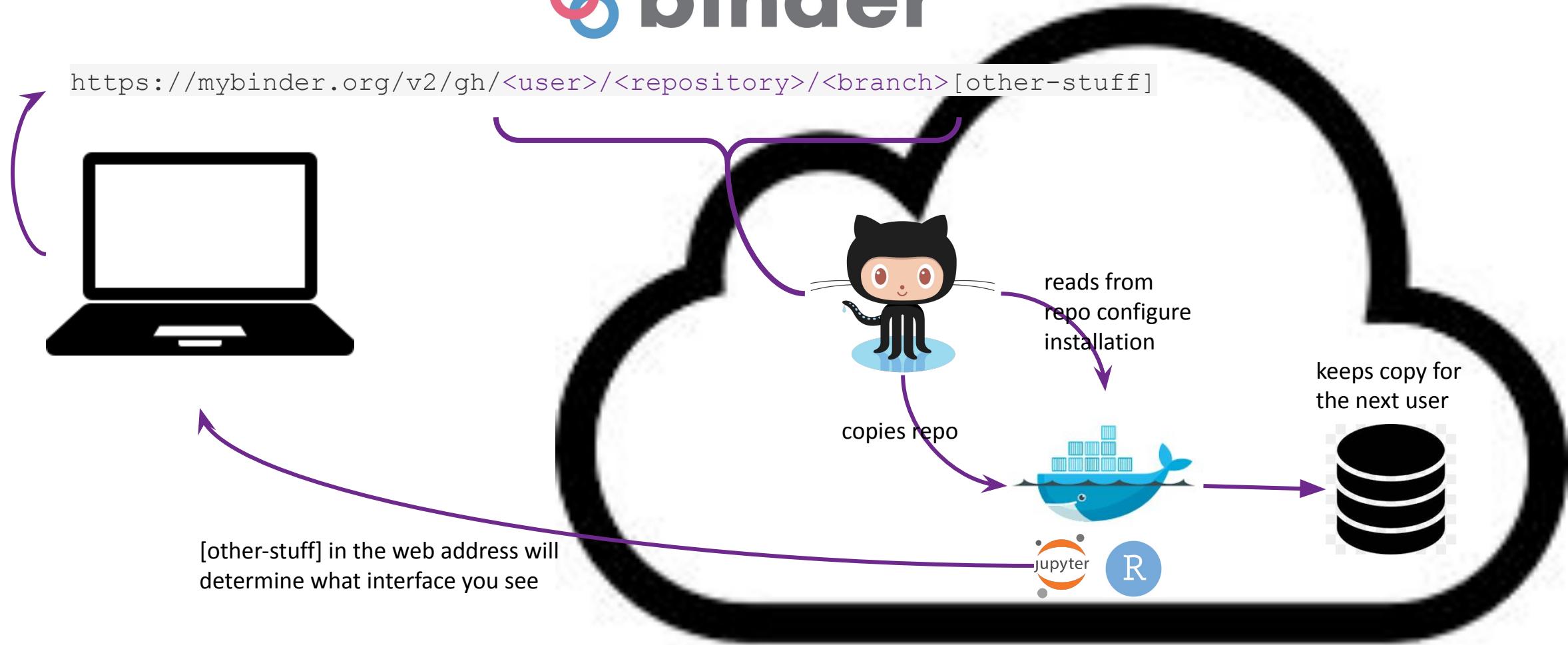
[building your own binder environments](#)

building your own containers (docker & singularity)

# Binder - repos runnable on the web



`https://mybinder.org/v2/gh/<user>/<repository>/<branch>[other-stuff]`



[other-stuff] in the web address will  
determine what interface you see

# Binder how - the files

## For python (jupyter)

- requirements.txt : tells binder what python packages to install
- runtime.txt : tells binder what version of python to install

see example in:

day1/example-python-repo

## For and R (with rstudio)

- runtime.txt: tells binder what version of R to use
- install.R: tells binder what R packages to install

see example in:

day1/example-r-repo

# Binder - how - the URL's

The standard format is:

`http://mybinder.org/v2/gh/<github-user>/<github-repo>/<branch>`

For this repo this is:

- <http://mybinder.org/v2/gh/krembilneuroinformatics/example-python-repo/HEAD>

But if directly to one particular python notebook. You could add the filepath to the end

?filepath=<filepath>

- Example:  
[https://mybinder.org/v2/gh/krembilneuroinformatics/example-python-repo.git/HEAD?filepath=example\\_notebook.ipynb](https://mybinder.org/v2/gh/krembilneuroinformatics/example-python-repo.git/HEAD?filepath=example_notebook.ipynb)

For the RStudio environment, we must add the following at the end of the URL: ?urlpath=rstudio

- Example:  
`http://mybinder.org/v2/gh/krembilneuroinformatics/example-r-repo/HEAD?urlpath=rstudio`

let's look at two example repo's in  
the kcni-school-lessons/day1

---

# Outline for this afternoon

Why reproducible science?

The things you need to know to get through this course:

Versioning and publishing code (github)

Versioning and publishing software (docker)

R with Rmarkdown (walk through)

Python in ipython notebooks (walk through)

-also google colab

**The fancy bits:**

building your own binder environments

**building your own containers (docker & singularity)**

# Step 1: write a Dockerfile

In an empty folder (new github repo)

Create a file named:

## Dockerfile

*Everything else in this folder will be copied inside the docker image.*

The Dockerfile contains the instructions for software installation. Commands are:

- **FROM** (at top line): points to another Docker image to start from
- **RUN**: will run an installation command
  - -run is followed by shell install commands
  - can add && to package lines into the same “layer”
- **ENTRYPOINT**: will determine one command that is run “by default”

```
FROM <base-image>

# stuff to install
RUN <installation command>

# more stuff (layer 2)
RUN <installation line1>
  && <installation line2>
```

Browse <https://hub.docker.com/> & kcni-school-lessons/envs for examples

# Step 1: write a Dockerfile

In an empty folder (new github repo)

Create a file named:

## Dockerfile

*Everything else in this folder will be copied inside the docker image.*

The Dockerfile contains the instructions for software installation. Commands are:

- **FROM** (at top line): points to another Docker image to start from
- **RUN**: will run an installation command
  - -run is followed by shell install commands
- **ENV**: can set environment variables
- **ENTRYPOINT**: will determine one command that is run “by default”

```
FROM rocker/verse:4.1.0

# adding plotting packages to from day 1 demo
RUN install2.r --error \
  --deps TRUE \
  ggrepel \
  ggthemes \
  here

## adding data grabbing packages
RUN apt-get update -qq \
  && apt-get -y --no-install-recommends \
  install wget curl git
```

Browse <https://hub.docker.com/> & kcni-school-lessosn/envs for examples

# Building your docker image

Use “docker build” to build and test your image on your local computer.

*Note “-t” is for tag - or the name you will give to your docker the last argument is the folder*

```
cd my_docker_folder/  
docker build -t my_new_docker ./
```

You can connect your github repo to dockerhub and have dockerhub re-build your container everytime you make a push to your repo



<https://hub.docker.com/>

# Running docker images

**To run your docker:**

```
docker run [options] <dockerhub_user>/<image>:<version>
```

example: `docker run [options] edickey/rstudio-school:latest`

**Important options when you run and image:**

- **--publish, -p**: allows port forwarding from the inside of the docker to the outside
  - we need this to connect rstudio or the inside of the docker to your computer's browser
- **--volume, -v**: use this to connect data outside the docker readable/writable by the docker software.
  - Docker can only “see” data that is connected to it (not everything on your computer)

# Running the kcnischool-rstudio image

**The full command to run the KCNI rstudio image is:**

```
docker run --rm -it \
-e DISABLE_AUTH=true \
-p 127.0.0.1:8787:8787 \
-v <path/to/your/data>:/home/rstudio/kcni-school-lessons \
edickie/kcnischool-rstudio:latest
```

**where** - `<path/to/my/data>` is the name of the folder on your computer where you cloned the `kjni-school-lessons` repo.

# docker-compose files save memory

We put a “docker-compose.yml” file at the base of the kcni-school-lessos repo. This file contains defaults of the docker options - so to run the docker you type

```
docker compose up rstudio
```

*Later in the week we will start jupyter with:*

```
docker compose up jupyter
```

```
version: '3'

services:
  rstudio:
    image: edickie/kcnischool-rstudio
    ports:
      - 8787:8787
    volumes:
      - ./:/home/rstudio/kcni-school-data
    environment:
      - DISABLE_AUTH=true

  jupyter:
    image: edickie/kcnischool-jupyter
    ports:
      - 8888:8888
    volumes:
      - ./:/home/neuro/kcni-school-data
```

# Docker vs Singularity

	Pro	Cons
Docker 	<ul style="list-style-type: none"><li>• Strong and building dev community</li><li>• can be pushed and pulled to dockerhub</li><li>• Works on any system (Windows, Mac, Linux)</li><li>• “Layers” decrease the hard disk space</li></ul>	<b>Needs “root” access</b>
Singularity 	<p><b>Does not need “root” access.</b></p> <p>Can converted from Docker</p>	<ul style="list-style-type: none"><li>• Smaller dev community</li><li>• Singularity hub exists, but is less used</li><li>• Only works on Linux</li></ul>

---

# Docker vs Singularity

take home...

- we write a Docker spec
- we run singularity containers on High performance computers and shared servers (like the CAMH SCC or SciNet)

---

# Translating Docker Usage to Singularity

	Docker	Singularity
Running	Docker run -it --rm	Singularity run
Mounting or binding a path	-v, --volume	-B, --bind
Attaching the “workdir”	-w, --workdir	-W
Port forwarding	-p, --publish	N/A
Removing the outside environment	N/A	-e, --cleanenv
Change the mount to \$HOME	N/A	-H, --home

---

# Building a singularity container from docker

## Direct from dockerhub to singularity

```
ssh <username>@teach.scinet.utoronto.ca

# note - in this example I am adding the new container to my $SCRATCH folder
mkdir $SCRATCH/test_sing_img

# singularity build <output-image>
docker://<dockerhubuser>/<dockernname>:<version>
singularity build singularity build \
    edickie_kcnischool-jupyter_latest-2021-07-02.sif \
    docker://edickie/kcnischool-jupyter:latest
```

---

## A note about “latest”

Most dockers will allow you to download the latest version of their software by typing “latest” instead of a version number

Problem - for reporting and debugging...you need the version number

So - for a real analysis - make sure to build a specific version and put the version number in the filename of the singularity image.

# Remember - many ways to engage



(during sessions)  
Use the chat or  
the ask question!



You can always return to the  
session and re-watch the videos  
after the session ends



come chat with us in KCNI  
Summer School Slack :)



virtually meet with us  
in gather.town



Tell us how the session went (post session survey):  
<https://forms.gle/ji18qLMZEZ9L16Ln6>



KCNI School@camh.ca

# Reproducibility crisis literature

7. Ioannidis JPA Why most published research findings are false. *PLoS Med.* 2, e124 (2005). This landmark paper outlines the ways in which common practices can lead to inflated levels of false positives.
8. Simmons JP, Nelson LD & Simonsohn U. False-positive psychology: undisclosed flexibility in data collection and analysis allows presenting anything as significant. *Psychol. Sci* 22, 1359–1366 (2011). This paper highlights the impact of common ‘questionable research practices’ on study outcomes and proposes a set of guidelines to prevent false-positive findings. [PubMed: 22006061]
9. Gelman A. & Loken E. The statistical crisis in science. *American Scientist* 102, 40 (2014).
10. Ioannidis JPA, Fanelli D, Dunne DD & Goodman SN Meta-research: evaluation and improvement of research methods and practices. *PLoS Biol.* 13, e1002264 (2015).
11. Collins FS & Tabak LA Policy: NIH plans to enhance reproducibility. *Nature* 505, 612–613 (2014). [PubMed: 24482835]