

Containers

QLS 612 - May 2023
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Fondation
Brain Canada
Foundation



HEALTHY BRAINS
HEALTHY LIVES



neuro
Montreal Neurological
Institute-Hospital



CONP
PCNO

Containers?

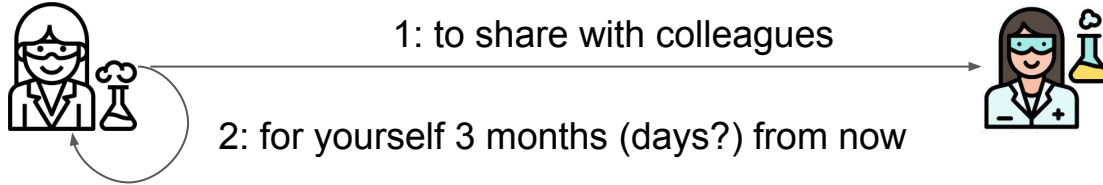


What we will talk about

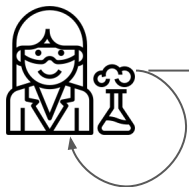
1. **Why** are containers useful for researchers
2. **Virtual Machines** (briefly)
3. Using and building containers with **Docker**
4. Using containers on supercomputers with **Singularity**

Why isolate environments

Document your software environment

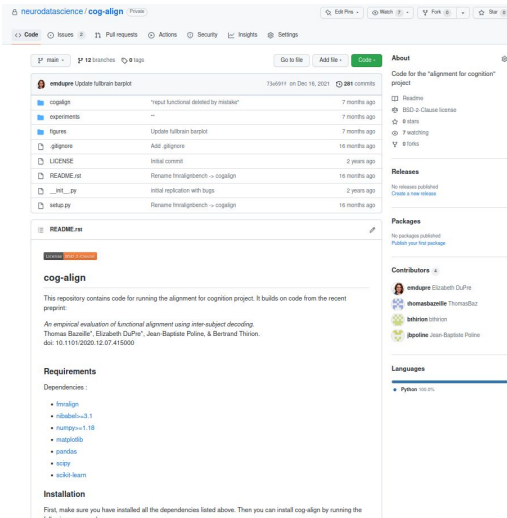


Document your software environment



1: to share with colleagues

2: for yourself 3 months (days?) from now



Requirements

Dependencies :

- **fmralign**
- **nibabel>=3.1**
- **numpy>=1.18**
- **matplotlib**
- **pandas**
- **scipy**
- **scikit-learn**

Installation

First, make sure you have installed all the dependencies listed above. Then you can install cog-align by running the following commands:

```
git clone https://github.com/neurodatascience/cog-align
cd cog-align
pip install -e .
```

Don't use the same environment for all projects



Changing dependencies may do unexpected things

A. Your updated the dependencies of an existing project

Don't use the same environment for all projects

Changing dependencies may do unexpected things

main C-PAC / requirements.txt

shnizzedy Merge develop-1.8.4 into
10 contributors

29 lines (29 sloc) 563 Bytes

```
1 boto3==1.7.37
2 click==6.7
3 configparser==3.7.4
4 future==0.16.0
5 git+https://git@github.com/FCP-IND1
6 lockfile==0.12.2
7 matplotlib==3.1.3
8 networkx==2.4
9 nibabel==2.3.3
10 nilearn==0.4.1
11 nipype==1.5.1
12 nose==1.3.7
13 numpy==1.21.0
```

master fmriprep / setup.cfg

mgxd rel(22.0.0rc3): bump minimum niworkflows [skip
6 contributors

127 lines (119 sloc) 2.72 KB

```
23 [options]
24 python_requires = >=3.7
25 install_requires =
26 nibabel >= 3.0
27 nipype >= 1.7.0, != 1.8.0
28 nitime
29 nitransforms >= 21.0.0
30 niworkflows >= 1.6.2
31 numpy
32 packaging
33 pandas
```

A. Your updated the dependencies of an existing project

B. Two projects use the same environment but need different versions of some dependencies

Do not install things into your system Python!

Common installation issues

Installing into the system Python on Linux

On Linux systems, a Python installation will typically be included as part of the distribution. Installing into this Python installation requires root access to the system, **and may interfere with the operation of the system package manager and other components** of the system if a component is unexpectedly upgraded using `pip`.

On such systems, it is often better to use a virtual environment or a per-user installation when installing packages with `pip`.

```
[surchs@marvin ~]$ which python
/usr/bin/python
[surchs@marvin ~]$ which pip
/usr/bin/pip
[surchs@marvin ~]$
```

Consider: a cake

Home · Cakes · Perfect Cream Cheese Pound Cake

Perfect Cream Cheese Pound Cake

Published by [Sally](#) on February 18, 2019 - [700 comments](#)



not sponsored, but I absolutely adore Nordic ware bundt pans.
10-12 cups of batter. [This one](#) is also gorgeous! 😊

- 5 **Bake:** Bake the cream cheese pound cake at 325°F (163°C). Hal the cake with aluminum foil to prevent over-browning.
- 6 **Cool, then invert:** Let the pound cool for about 2 hours in the plate and cool completely before serving.

Consider: a cake

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Isolate environments to handle different requirements



Why not just python virtual environments?

External Dependencies

fMRIPrep is written using Python 3.8 (or above), and is based on [nipy](#).

fMRIPrep requires some other neuroimaging software tools that are not handled by the Python's packaging system (Pypi) used to deploy the `fmripred` package:

- [FSL](#) (version 6.0.5.1)
- [ANTs](#) (version 2.3.3 - NeuroDocker build)
- [AFNI](#) (version 22.3.06)
- [C3D](#) (version 1.3.0)
- [FreeSurfer](#) (version 7.3.2)
- [ICA-AROMA](#) (version 0.4.5)
- [bids-validator](#) (version 1.8.0)
- [connectome-workbench](#) (version 1.5.0)

- Not all binary dependencies are on Anaconda
- Not everything is written in Python or R
- Your Operating System (**OS**) also has packages and a package manager
- The same version problems apply to these

Need to capture the (entire) compute environment



1: to share with colleagues



2: for yourself 3 months (days?) from now

Virtual Machines

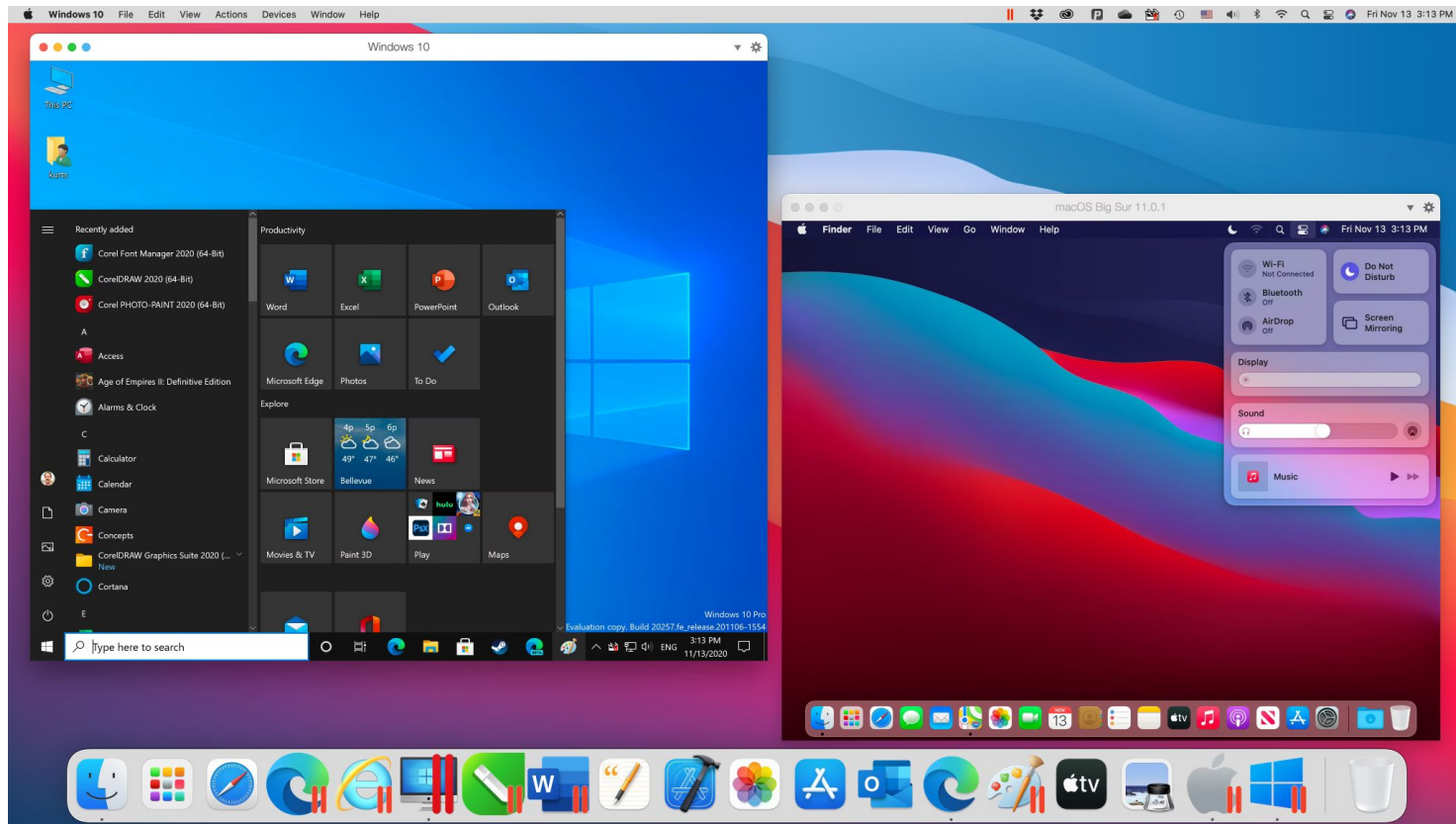
How can we isolate different OS environments?



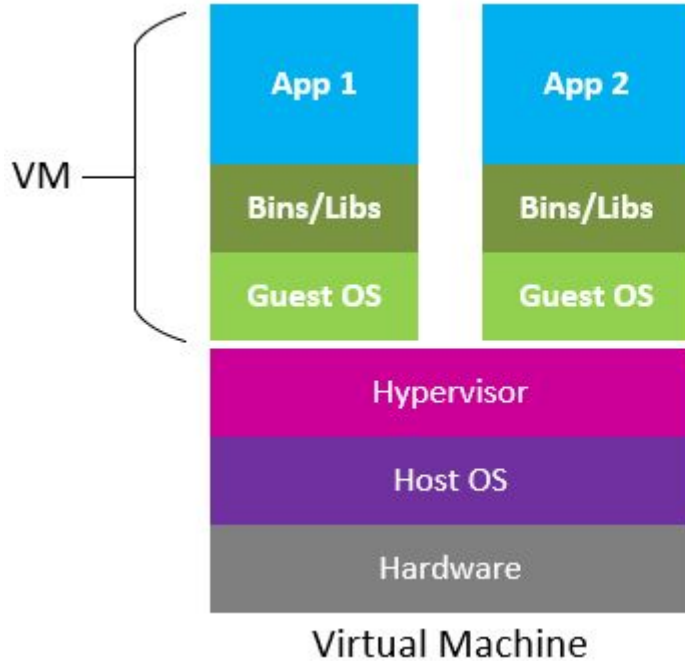
Buy a lot of computers?



Virtual Machines: Let's simulate the computers



Virtual Machine (VM)



What is it?

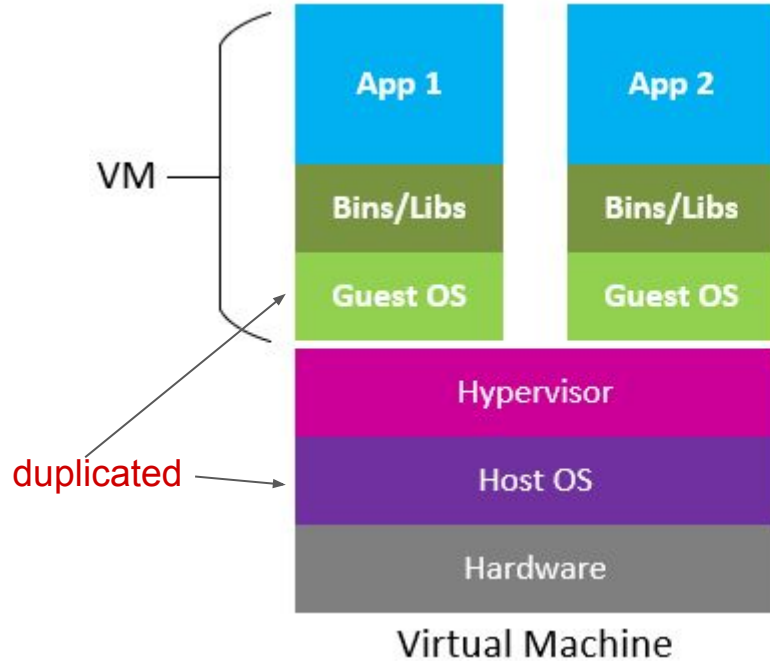
- entire computer hardware simulated (**hardware virtualization**)
- you can choose what hardware is exposed (CPU, RAM, hard-drive, ...)
- the VM is fully isolated from the Host OS: you get to install any OS inside the VM

How do I make one?

Get Hypervisor / Virtualization software

- VirtualBox (partly OSS)
- VMWare (commercial)
- Hyper-V (Windows)
- ...

Virtual Machine (VM)



Good:

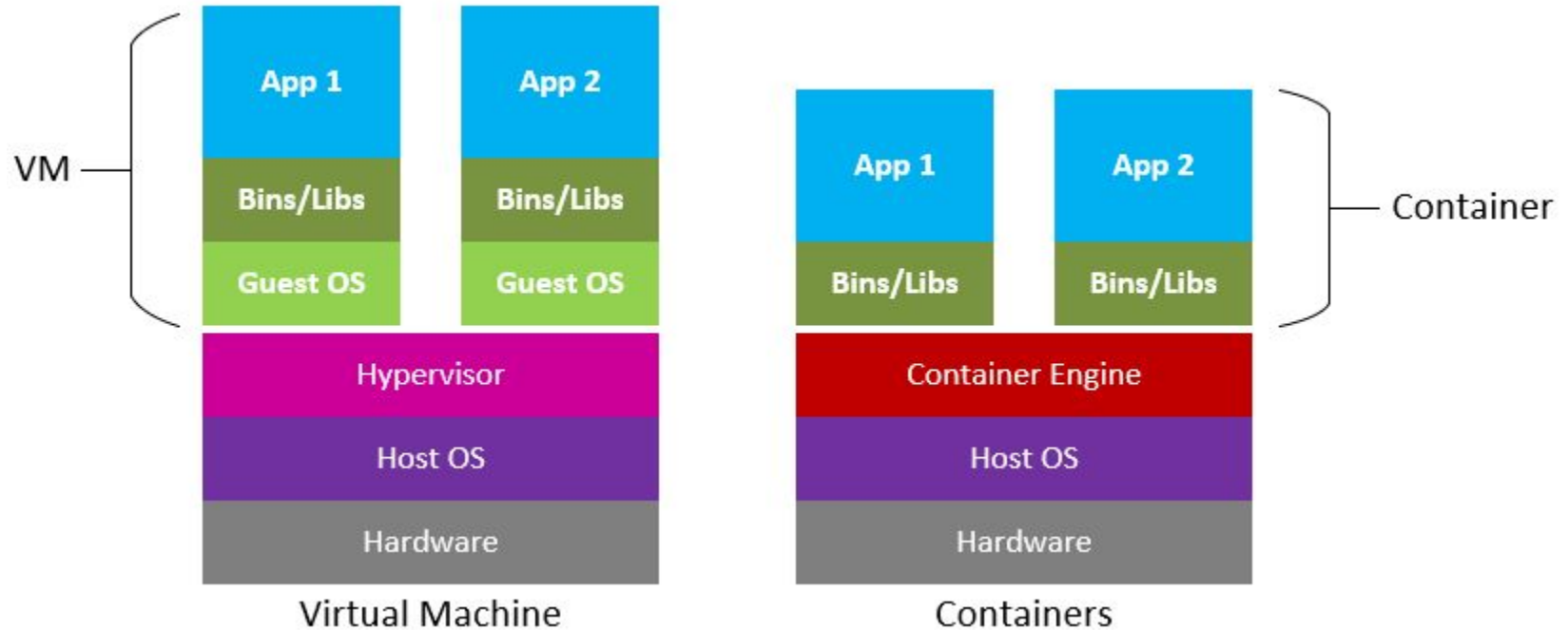
- can run anything a computer runs (Windows, Linux, Hackintosh)
- can make a snapshot to share with others (Neurodebian used to have a VM)
- good way to test things across many systems
- full system isolation

But:

- doesn't share resources with host -> BIG
- slow to start up, stop, resume
- cumbersome to configure for each project
- duplicates things (every VM needs OS / Kernel)
- no easy way to "get" and use VMs from other people
- you can't use it on a supercomputer

Containers

Containers: let's all share the same kernel, but in boxes



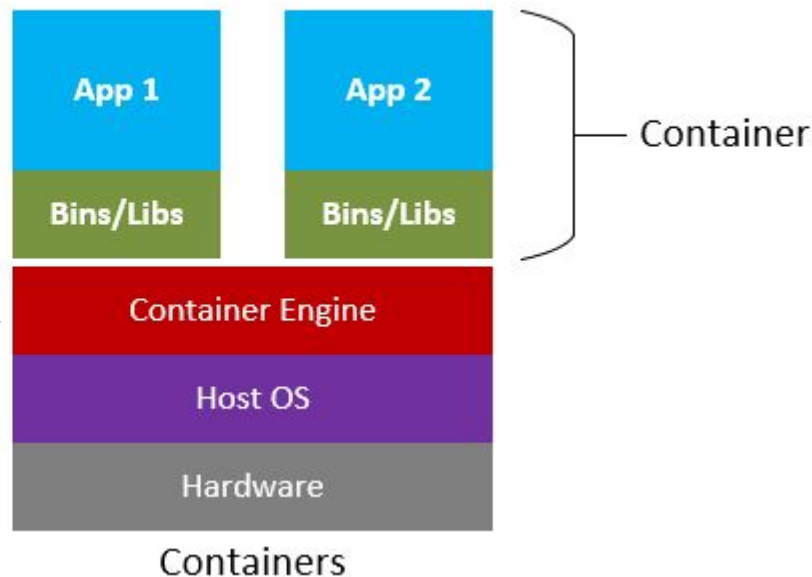
Containers

What are they

- isolated environments sharing the same kernel / OS -> (**OS virtualization**)
- from the inside, a container looks like a separate computer, can't see outside
- within each container, you can have your desired binary and library dependencies

How do I make one

- use a container implementation
- **docker** is the most widely used
- Singularity is used on supercomputers



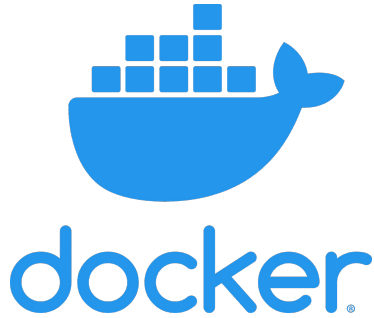
Virtual Machine



Container



What is Docker



Docker Engine

- a command line program
- gets and builds Docker images and runs Docker containers



Docker Hub

- a website / web service
- a central repository to store and share Docker container images (commercial)
- other container image registries exist

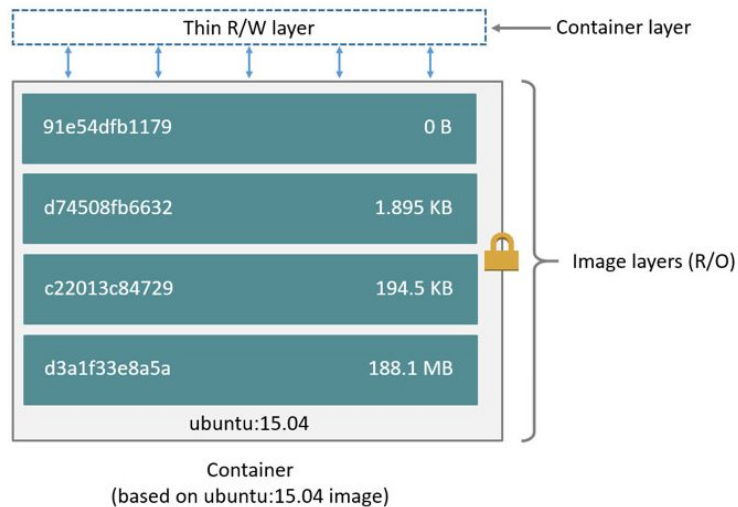
Docker **image** and **container**: what's the difference

Docker image

- a **read-only** snapshot of an environment
- organized in **layers**
- changing an image adds more layers
- can be stored on Dockerhub or as a file
- images can share identical layers
- can make your own with a **Dockerfile**

Docker container

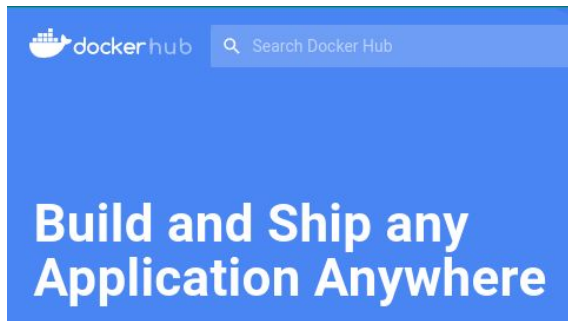
- a **live instance** of a Docker image
- has a thin writable layer that dies with it
- **one image** can spawn **many containers**



How can I get my own Docker container going

Use an **existing image**

- Dockerhub: a repository of docker images
<https://hub.docker.com/>
- You can pull an image with `docker pull`



Build your **own image**

- a `Dockerfile` lets you describe the exact image you want to create
- Start from one image and add to it



```
# syntax=docker/dockerfile:1
FROM ubuntu:18.04
LABEL org.opencontainers.image.authors="org@example.com"
COPY . /app
RUN make /app
RUN rm -r $HOME/.cache
CMD python /app/app.py
```

Let's look at both

Exercise 1: Run a container from Dockerhub

- Find an image we like: https://hub.docker.com/_/hello-world
- Take a look at it
- Pull the image
- Run the image

Exercise 1: Run a container from Dockerhub

- Find an image we like: https://hub.docker.com/_/hello-world
- Take a look at it
- Pull the image
- Run the image

Copy and paste to pull this image

```
docker pull hello-world
```



[View Available Tags](#)

```
surchs@deeptthought:~/Documents/docker_place
Status: Downloaded newer image for hello-world:latest

Hello from Docker!
This message shows that your installation appears to be working correctly.

To generate this message, Docker took the following steps:
1. The Docker client contacted the Docker daemon.
2. The Docker daemon pulled the "hello-world" image from the Docker Hub.
   (amd64)
3. The Docker daemon created a new container from that image which runs the
```

Exercise 1: Run a container from Dockerhub

Summary

- Dockerhub has images
- Image tags are important
- We can
 - retrieve images with `docker pull`
 - create and immediately run a container from an image with `docker run`

Let's find a more useful image

Exercise 2: work with a container that has conda

I don't have conda installed on my system. But there is a Docker image. Let's try!

- Find a Docker image: <https://hub.docker.com/r/continuumio/miniconda3/>
- Pick a tag, then pull the image `$ docker pull continuumio/miniconda3:22.11.1-alpine`
- Run it

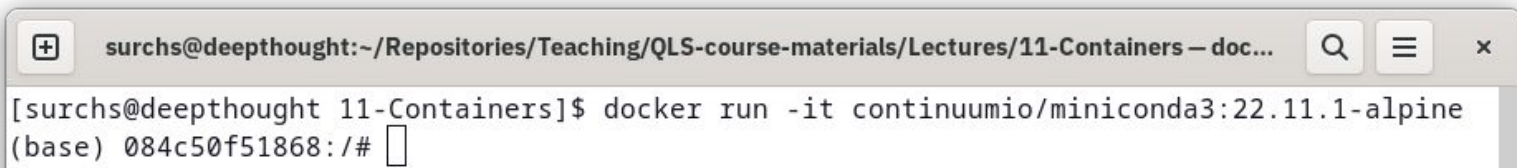
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- Pick a tag, then pull the image `$ docker pull continuumio/miniconda3:22.11.1-alpine`
- Run it

Oh, nothing happened?

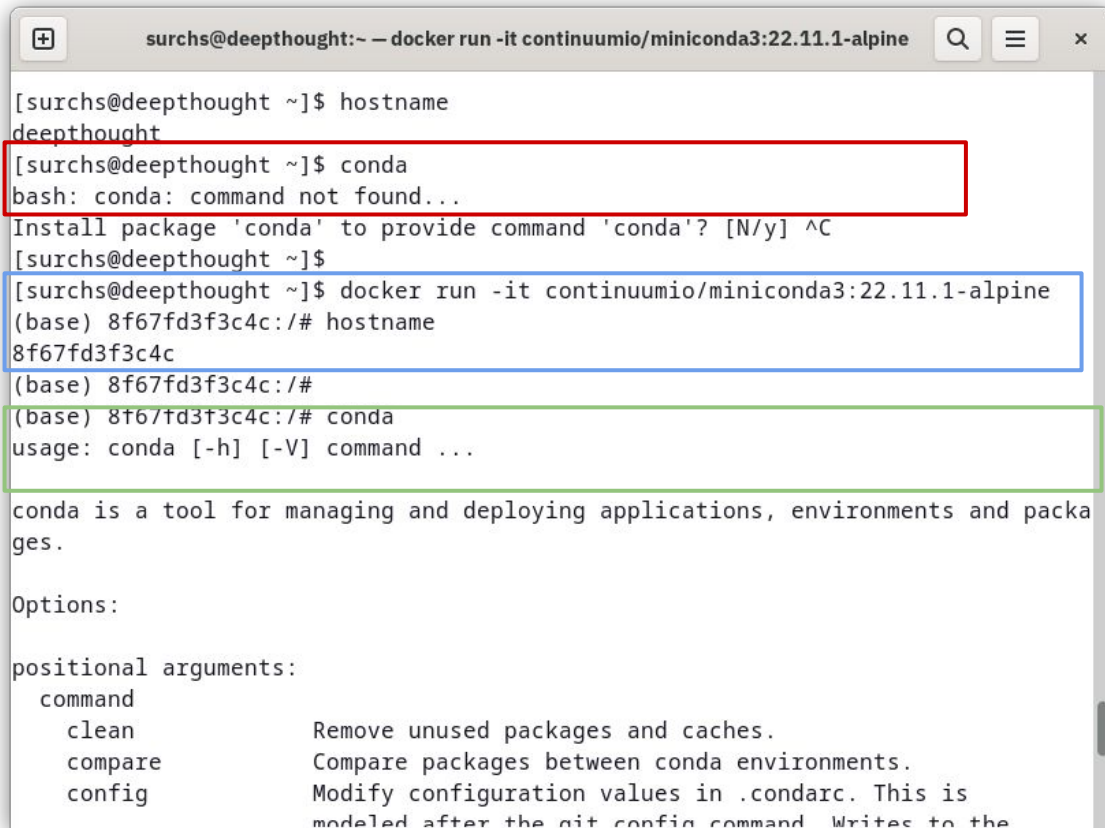
- Let's run it interactively to take a look inside

A terminal window with a title bar that reads "surchs@deeptthought:~/Repositories/Teaching/QLS-course-materials/Lectures/11-Containers – doc...". The terminal shows the command "docker run -it continuumio/miniconda3:22.11.1-alpine" being executed. The prompt changes to "(base) 084c50f51868:/#" and a cursor is visible at the end of the line.

```
surchs@deeptthought:~/Repositories/Teaching/QLS-course-materials/Lectures/11-Containers – doc...  
[surchs@deeptthought 11-Containers]$ docker run -it continuumio/miniconda3:22.11.1-alpine  
(base) 084c50f51868:/#
```

Looking around inside a container

- no conda on my machine
- starting container changes the look of my terminal and the name of my computer
- inside of the container I have access to conda



```
surchs@deepthought:~ -- docker run -it continuumio/miniconda3:22.11.1-alpine

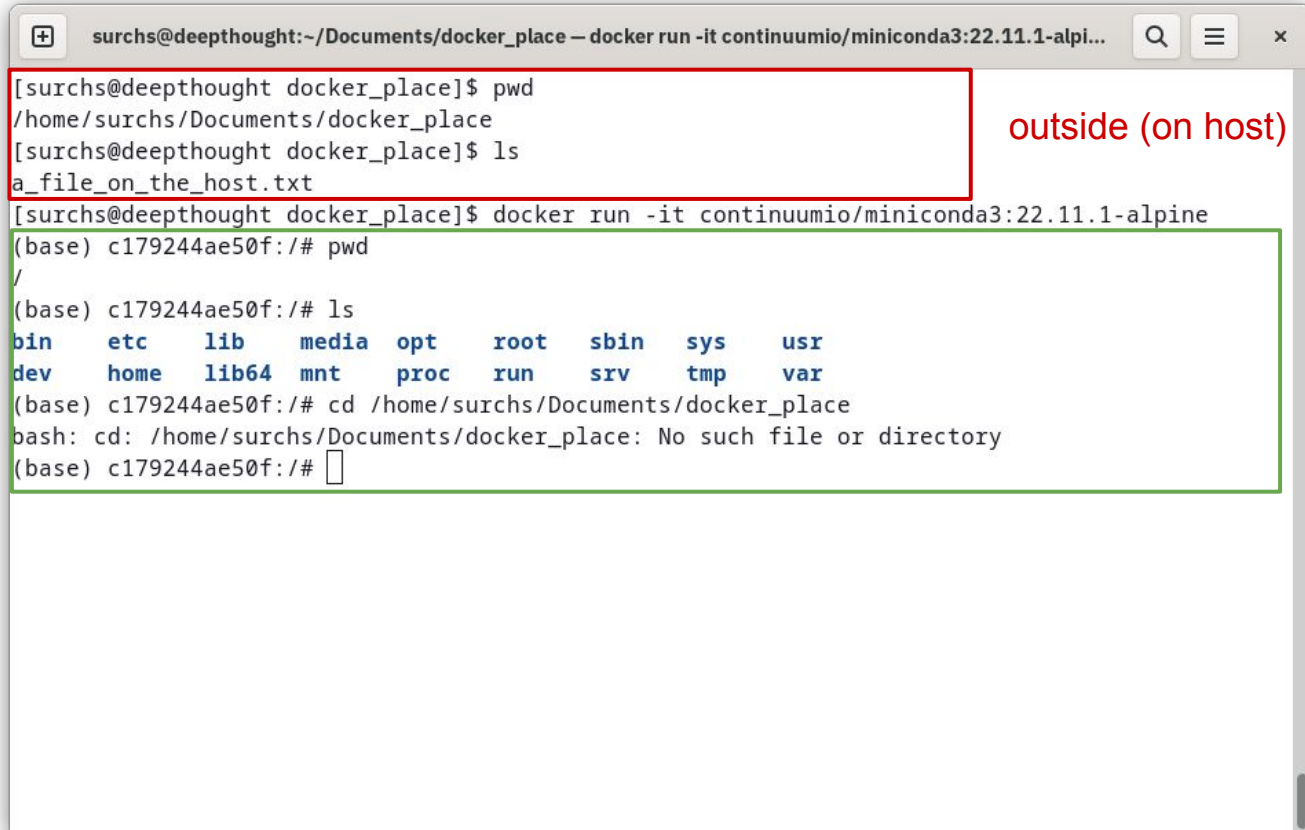
[surchs@deepthought ~]$ hostname
deepthought
[surchs@deepthought ~]$ conda
bash: conda: command not found...
Install package 'conda' to provide command 'conda'? [N/y] ^C
[surchs@deepthought ~]$
[surchs@deepthought ~]$ docker run -it continuumio/miniconda3:22.11.1-alpine
(base) 8f67fd3f3c4c:/# hostname
8f67fd3f3c4c
(base) 8f67fd3f3c4c:/#
(base) 8f67fd3f3c4c:/# conda
usage: conda [-h] [-V] command ...

conda is a tool for managing and deploying applications, environments and packages.

Options:

positional arguments:
  command
    clean                Remove unused packages and caches.
    compare              Compare packages between conda environments.
    config              Modify configuration values in .condarc. This is
                        modeled after the git config command. Writes to the
```

By default the container doesn't see files on the host ...



The image shows a terminal window with a title bar that reads "surchs@deeptthought:~/Documents/docker_place — docker run -it continuumio/miniconda3:22.11.1-alpi...". The terminal content is divided into two sections by colored borders. The top section, outlined in red, shows the host's perspective: a user named "surchs" runs "pwd" and gets "/home/surchs/Documents/docker_place", then runs "ls" and lists "a_file_on_the_host.txt". To the right of this section is the red text "outside (on host)". The bottom section, outlined in green, shows the container's perspective: a user runs "docker run -it continuumio/miniconda3:22.11.1-alpine", then "pwd" and gets "/", then "ls" and lists standard Linux directories. To the right of this section is the green text "inside (in container)". The bottom section ends with an attempt to "cd /home/surchs/Documents/docker_place", which fails with the message "bash: cd: /home/surchs/Documents/docker_place: No such file or directory".

```
[surchs@deeptthought docker_place]$ pwd
/home/surchs/Documents/docker_place
[surchs@deeptthought docker_place]$ ls
a_file_on_the_host.txt

[surchs@deeptthought docker_place]$ docker run -it continuumio/miniconda3:22.11.1-alpine
(base) c179244ae50f:/# pwd
/
(base) c179244ae50f:/# ls
bin    etc    lib    media  opt    root   sbin   sys    usr
dev    home   lib64  mnt     proc   run    srv    tmp    var
(base) c179244ae50f:/# cd /home/surchs/Documents/docker_place
bash: cd: /home/surchs/Documents/docker_place: No such file or directory
(base) c179244ae50f:/#
```


... and the host can't see files on the container ...

```
surchs@deepthought:~/Documents/docker_place
[surchs@deepthought docker_place]$ docker run -it continuumio/miniconda3:22.11.1-alpine
(base) fa973dee589c:/# ls
bin      etc      lib      media    opt      root     sbin     sys      usr
dev      home    lib64    mnt      proc     run      srv      tmp      var
(base) fa973dee589c:/# touch container_file.txt
(base) fa973dee589c:/# ls
bin              lib              proc              sys
container_file.txt lib64             root              tmp
dev              media            run               usr
etc              mnt              sbin              var
home             opt              srv
(base) fa973dee589c:/#
exit
[surchs@deepthought docker_place]$ ls /container_file.txt
ls: cannot access '/container_file.txt': No such file or directory
[surchs@deepthought docker_place]$
```

inside
(in container)

outside (on host)

... and files written to a container are tied to it!

```
surchs@deephthought:~/Documents/docker_place
[surchs@deephthought docker_place]$ docker run -it continuumio/miniconda3:22.11.1-alpine
(base) 0fb93cb1903e:/# ls
bin      etc      lib      media   opt      root     sbin     sys      usr
dev      home    lib64    mnt     proc     run      srv      tmp      var
(base) 0fb93cb1903e:/# touch file1.txt
(base) 0fb93cb1903e:/# ls
bin      file1.txt  lib64     opt      run      sys      var
dev      home      media     proc     sbin     tmp
etc      lib       mnt       root     srv      usr
(base) 0fb93cb1903e:/#
exit
[surchs@deephthought docker_place]$ docker run -it continuumio/miniconda3:22.11.1-alpine
(base) 2b690713ac84:/# ls
bin      etc      lib      media   opt      root     sbin     sys      usr
dev      home    lib64    mnt     proc     run      srv      tmp      var
(base) 2b690713ac84:/# ls file1.txt
ls: file1.txt: No such file or directory
(base) 2b690713ac84:/#
exit
[surchs@deephthought docker_place]$
```

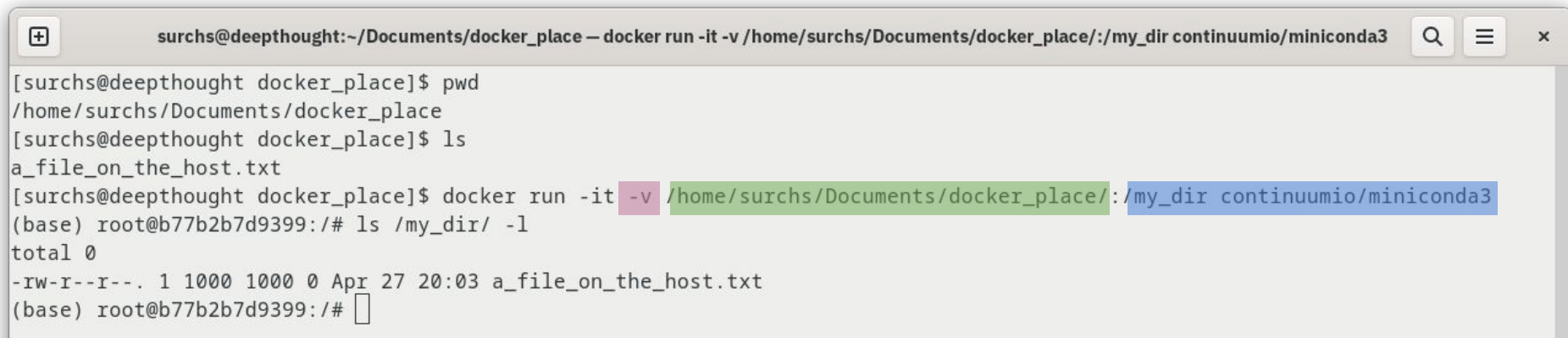
first container instance

second container instance

-> Don't write anything (worth keeping) to a container

How do we share files between host and container?

- Bind-mount a path on the host to a path in the container



```
surchs@deepthought:~/Documents/docker_place — docker run -it -v /home/surchs/Documents/docker_place:/my_dir continuumio/miniconda3
[surchs@deepthought docker_place]$ pwd
/home/surchs/Documents/docker_place
[surchs@deepthought docker_place]$ ls
a_file_on_the_host.txt
[surchs@deepthought docker_place]$ docker run -it -v /home/surchs/Documents/docker_place:/my_dir continuumio/miniconda3
(base) root@b77b2b7d9399:/# ls /my_dir/ -l
total 0
-rw-r--r--. 1 1000 1000 0 Apr 27 20:03 a_file_on_the_host.txt
(base) root@b77b2b7d9399:/#
```

- The bind-mount will be created in the container, even if it exists already!
- Make sure you provide a path (starts with / or ./) - or use --mount

Exercise 2: work with a container that has conda

Summary

- We can connect to an interactive shell in a container with `docker run -it`
- By default the container cannot see or write the filesystem of the host
- We can “mount” a path on the host into the container with

```
docker run -v /host/path:/container/path OR
```

```
docker run --mount type=bind,source=/host/path,target=/container/path
```

- It's a bad idea to write into the container directly

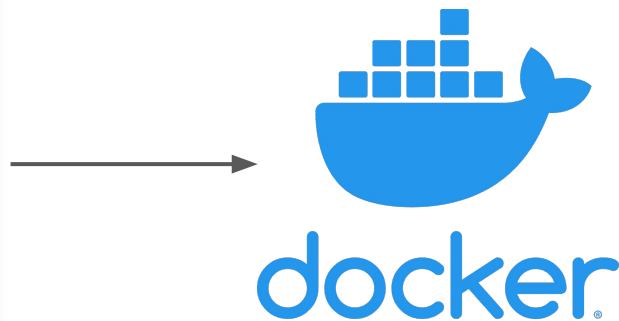
So how do I make persistent changes to the image?

Exercise 3: Containerize a classifier

- I want to run the classification script from 08-Machine_Learning_1
- but I did not setup all of the requirements to run it

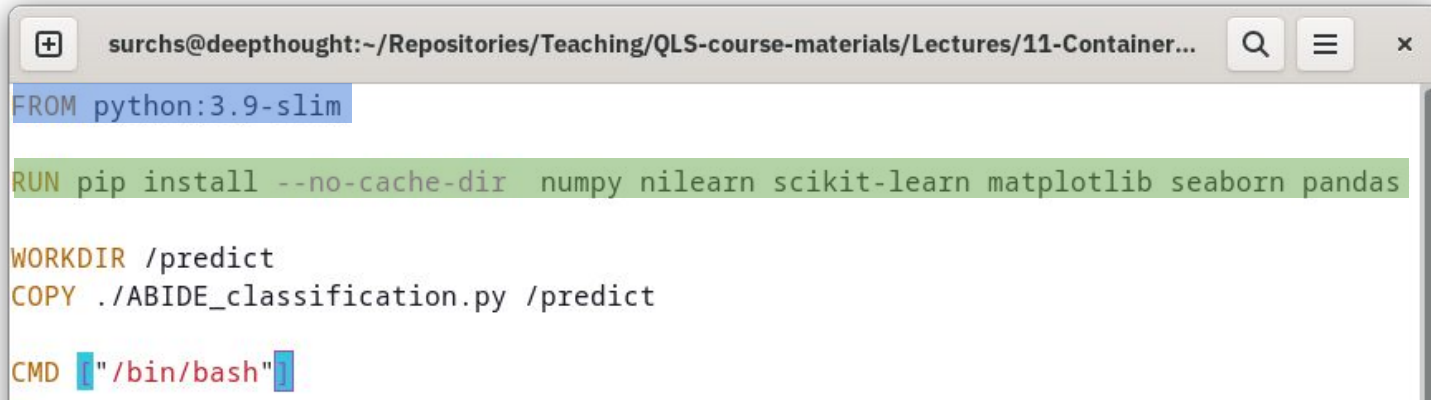
```
surchs@deeptthought:~/Repositories/Teaching/QLS-course-materials/Lectures/11-Co...
File: ABIDE_classification.py

1  ## Imports
2  from nilearn import datasets
3  from nilearn.connectome import ConnectivityMeasure
4  import pandas as pd
5  import numpy as np
6  import argparse
7  from sklearn import preprocessing
8  from sklearn.model_selection import train_test_split
9  from sklearn.linear_model import LogisticRegression
10 from sklearn.ensemble import RandomForestClassifier
11 from sklearn.metrics import precision_recall_fscore_support
12 from sklearn.metrics import confusion_matrix
13
14
15 def extract_connectome_features(func_data, measure):
16     ''' A function to calculate connectome based on timeseries data an
17         d similarity measure
18         '''
19     connectome_matrix = measure.fit_transform([func_data])[0]
20     tril_idx = np.tril_indices(len(connectome_matrix), k=-1)
```



Exercise 3: Containerize a classifier

1. Start **from** an existing image
2. Define **changes** on top of that with a **Dockerfile**
3. Build a new image using the **docker build** command

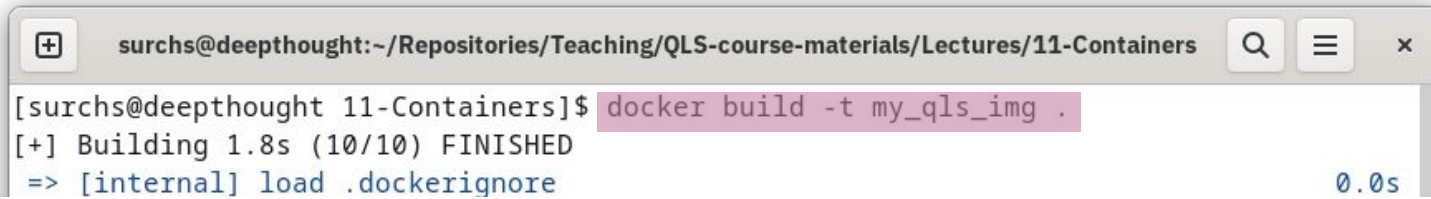


```
FROM python:3.9-slim

RUN pip install --no-cache-dir numpy nilearn scikit-learn matplotlib seaborn pandas

WORKDIR /predict
COPY ./ABIDE_classification.py /predict

CMD ["/bin/bash"]
```



```
[surchs@deepthought 11-Containers]$ docker build -t my_qls_img .
[+] Building 1.8s (10/10) FINISHED
=> [internal] load .dockerignore                                0.0s
```

Exercise 3: Containerize a classifier

4. We have **built our own docker image and stored it locally** on the computer

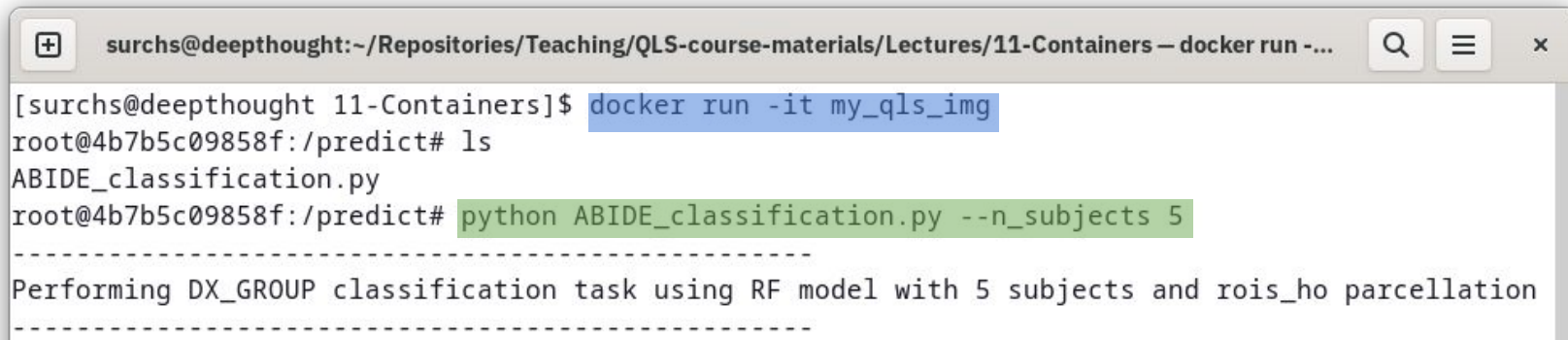
```
surchs@deephought:~/Repositories/Teaching/QLS-course-materials/Lectures/11-Containers
[surchs@deephought 11-Containers]$ docker build -t my_qls_img .
[+] Building 1.8s (10/10) FINISHED
=> [internal] load .dockerignore                                0.0s
```

```
surchs@deephought:~/Repositories/Teaching/QLS-course-materials/Lectures/11-Containers
[surchs@deephought 11-Containers]$ docker images
```

REPOSITORY	TAG	IMAGE ID	CREATED	SIZE
my_qls_img	latest	5ed020af550d	6 minutes ago	650MB

Run the classifier script

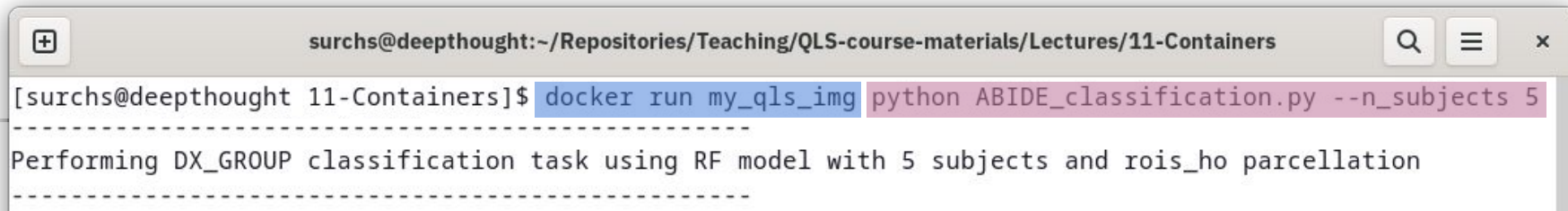
1. **Run the container and attach** an interactive shell
2. Inside the container, **run the classification script** like we would on the host



A terminal window titled "surchs@deeptthought:~/Repositories/Teaching/QLS-course-materials/Lectures/11-Containers — docker run -...". The prompt is "[surchs@deeptthought 11-Containers]\$". The first command is "docker run -it my_qls_img", which is highlighted in blue. The prompt changes to "root@4b7b5c09858f:/predict#". The second command is "ls", showing "ABIDE_classification.py". The third command is "python ABIDE_classification.py --n_subjects 5", highlighted in green. The output is "Performing DX_GROUP classification task using RF model with 5 subjects and rois_ho parcellation", preceded and followed by dashed lines.

```
[surchs@deeptthought 11-Containers]$ docker run -it my_qls_img
root@4b7b5c09858f:/predict# ls
ABIDE_classification.py
root@4b7b5c09858f:/predict# python ABIDE_classification.py --n_subjects 5
-----
Performing DX_GROUP classification task using RF model with 5 subjects and rois_ho parcellation
-----
```

3. We can also **run the container** and then **run the script as an argument**

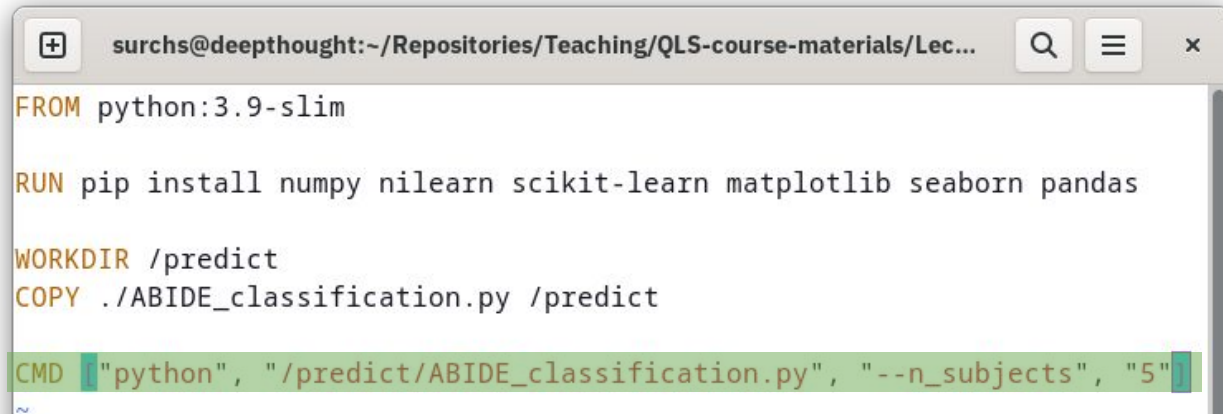


A terminal window titled "surchs@deeptthought:~/Repositories/Teaching/QLS-course-materials/Lectures/11-Containers". The prompt is "[surchs@deeptthought 11-Containers]\$". The command is "docker run my_qls_img python ABIDE_classification.py --n_subjects 5", with "docker run my_qls_img" highlighted in blue and the rest in purple. The output is "Performing DX_GROUP classification task using RF model with 5 subjects and rois_ho parcellation", preceded and followed by dashed lines.

```
[surchs@deeptthought 11-Containers]$ docker run my_qls_img python ABIDE_classification.py --n_subjects 5
-----
Performing DX_GROUP classification task using RF model with 5 subjects and rois_ho parcellation
-----
```


I keep writing the same command, what if I didn't have to?

1. The **CMD keyword in the Dockerfile** defines a default command
2. Unless the user provides their own command, the default CMD is run

A screenshot of a code editor window showing a Dockerfile. The window title is "surchs@deeptthought:~/Repositories/Teaching/QLS-course-materials/Lec...". The Dockerfile content is as follows:

```
FROM python:3.9-slim

RUN pip install numpy nilearn scikit-learn matplotlib seaborn pandas

WORKDIR /predict
COPY ./ABIDE_classification.py /predict

CMD ["python", "/predict/ABIDE_classification.py", "--n_subjects", "5"]
```

The last line, the CMD instruction, is highlighted with a green background.

3. To reflect the changes in the Dockerfile, I have to **rebuild** my image

Exercise 3: Containerize a classifier

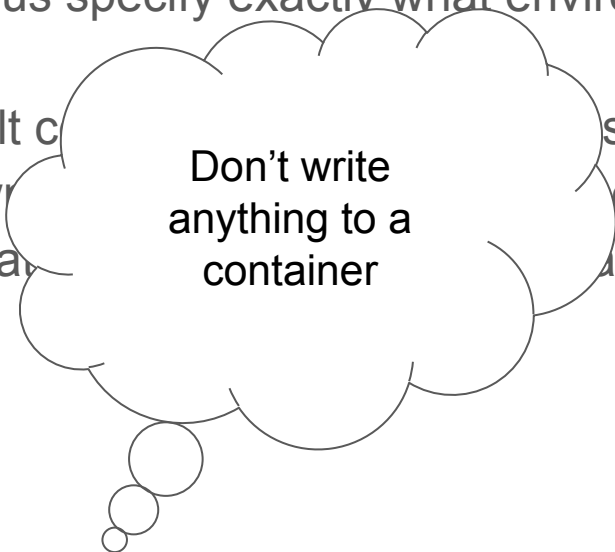
First summary:

- A `Dockerfile` lets us specify exactly what environment we want in a container
- `CMD` defines a default command that runs if the user doesn't provide one
- We can build our own docker image on top of an existing one
- `docker build` creates the specified docker image on our host computer

Exercise 3: Containerize a classifier

First summary:

- A `Dockerfile` lets us specify exactly what environment we want in a container
 - `CMD` defines a default command
 - We can build our own image
 - `docker build` creates a new image
- ... if the user doesn't provide one
... using an existing one
... stored on our host computer



Every time I run a new container, I have to download my data again ...

Loading or storing data in a container is not great



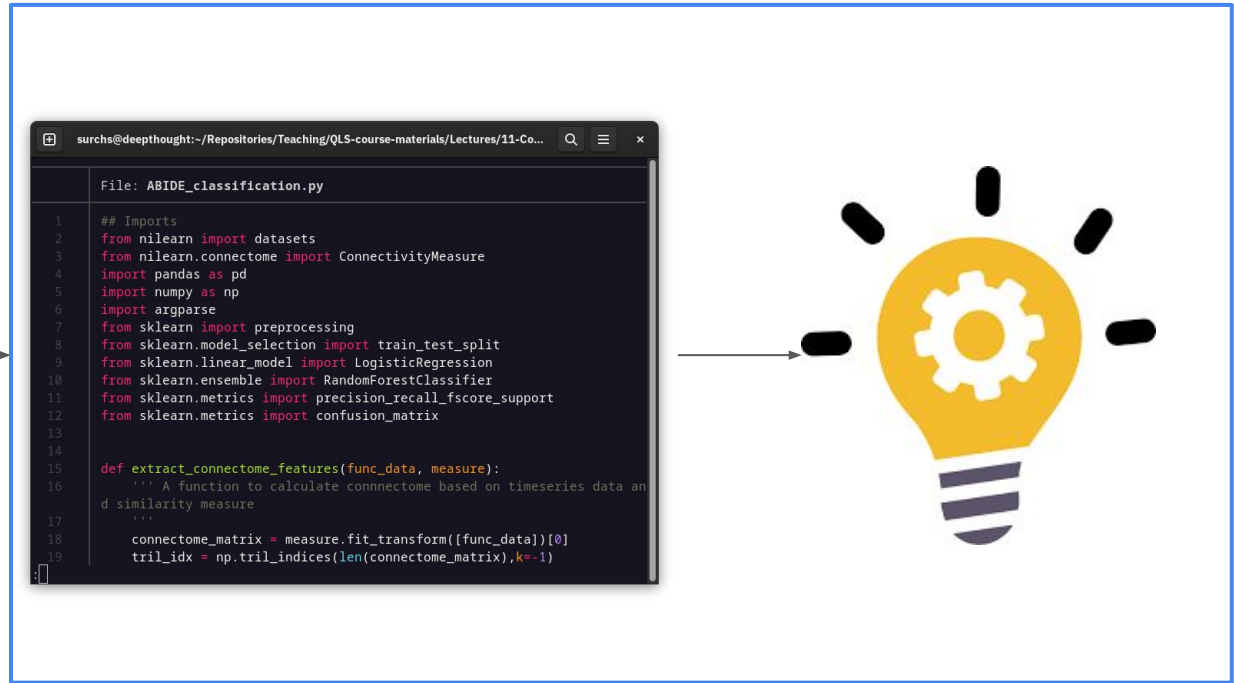
```
File: ABIDE_classification.py

1  ## Imports
2  from nilearn import datasets
3  from nilearn.connectome import ConnectivityMeasure
4  import pandas as pd
5  import numpy as np
6  import argparse
7  from sklearn import preprocessing
8  from sklearn.model_selection import train_test_split
9  from sklearn.linear_model import LogisticRegression
10 from sklearn.ensemble import RandomForestClassifier
11 from sklearn.metrics import precision_recall_fscore_support
12 from sklearn.metrics import confusion_matrix
13
14
15 def extract_connectome_features(func_data, measure):
16     ''' A function to calculate connectome based on timeseries data and
17     similarity measure
18     '''
19     connectome_matrix = measure.fit_transform([func_data])[0]
20     tril_idx = np.tril_indices(len(connectome_matrix), k=-1)
```



not persistent, not shareable with other container, slow, ...

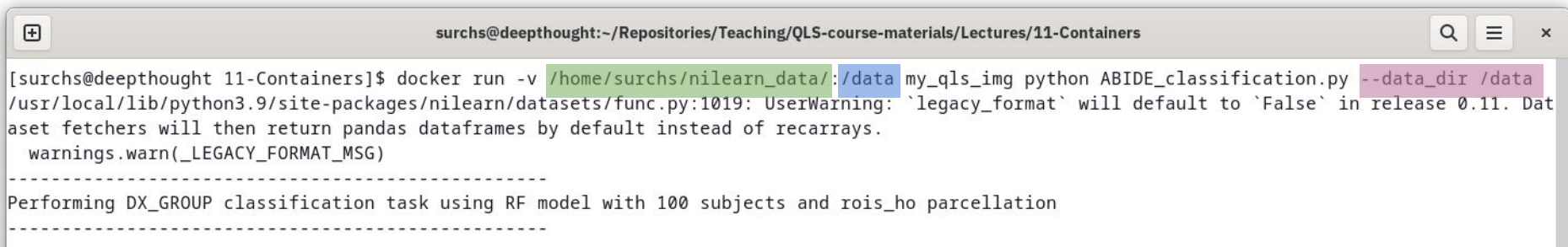
Containers are for processes, not data



-> let's bind-mount the data into the container from our host machine

Provide the downloaded data via bind-mount

1. Bind the **path to the data on your computer** to a **path in the container**
2. Provide the path in the container **to the classifier script**
3. The script inside the container only sees the mounted path

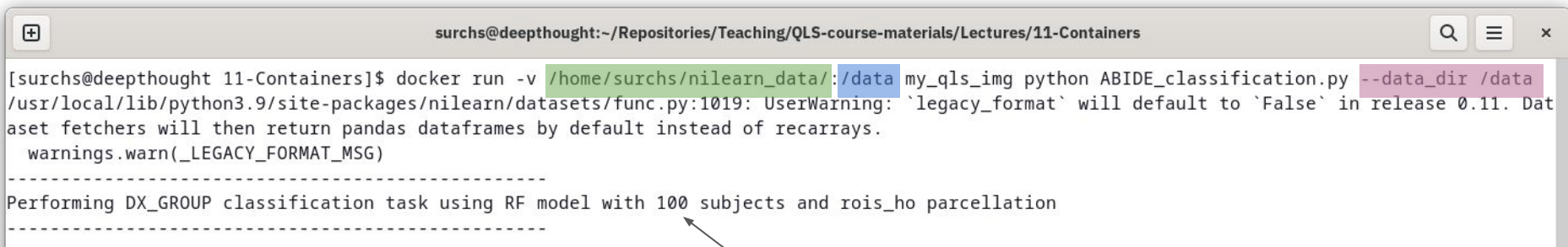


```
surchs@deeptthought:~/Repositories/Teaching/QLS-course-materials/Lectures/11-Containers

[surchs@deeptthought 11-Containers]$ docker run -v /home/surchs/nilearn_data:/data my_qls_img python ABIDE_classification.py --data_dir /data
/usr/local/lib/python3.9/site-packages/nilearn/datasets/func.py:1019: UserWarning: `legacy_format` will default to `False` in release 0.11. Dataset fetchers will then return pandas dataframes by default instead of recarrays.
  warnings.warn(_LEGACY_FORMAT_MSG)
-----
Performing DX_GROUP classification task using RF model with 100 subjects and rois_ho parcellation
-----
```

Provide the downloaded data via bind-mount

1. Bind the **path to the data on your computer** to a **path in the container**
2. Provide the path in the container **to the classifier script** as a parameter
3. The script inside the container only sees the mounted path



```
surchs@deeptthought:~/Repositories/Teaching/QLS-course-materials/Lectures/11-Containers

[surchs@deeptthought 11-Containers]$ docker run -v /home/surchs/nilearn_data:/data my_qls_img python ABIDE_classification.py --data_dir /data
/usr/local/lib/python3.9/site-packages/nilearn/datasets/func.py:1019: UserWarning: `legacy_format` will default to `False` in release 0.11. Dat
aset fetchers will then return pandas dataframes by default instead of recarrays.
  warnings.warn(_LEGACY_FORMAT_MSG)
-----
Performing DX_GROUP classification task using RF model with 100 subjects and rois_ho parcellation
-----
```

Bonus: why is the classifier running with 100 subjects now?

Exercise 3: Containerize a classifier

- We have a `Dockerfile` that describes our desired environment
- We have built a docker image that contains all dependencies and the code
- We have a default command so a user can just run a container
- -> **We have containerized the classifier**



Exercise 3: Containerize a classifier

- We have a `Dockerfile` that describes our desired environment
- We have built a docker image that contains all dependencies and the code
- We have a default command so a user can just run a container
- -> **We have containerized the classifier**

What if I'd like to make a change to the code?

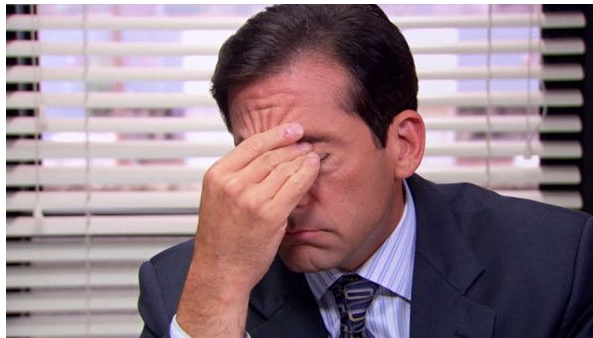


Exercise 3: Containerize a classifier

- We have a `Dockerfile` that describes our desired environment
- We have built a docker image that contains all dependencies and the code
- We have a default command so a user can just run a container
- -> **We have containerized the classifier**

What if I'd like to make a change to the code?

```
$ docker build  
$ docker build  
$ docker build  
$ docker build  
$ docker build  
$ docker build .
```



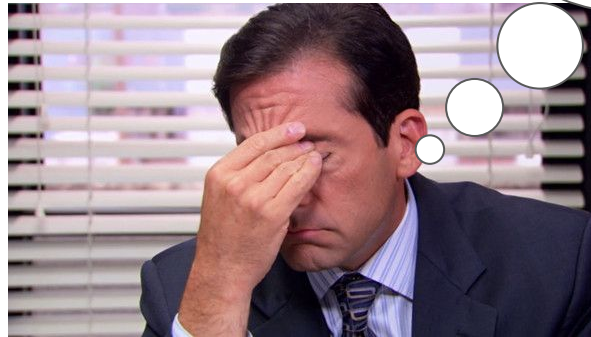
Exercise 3: Containerize a classifier

- We have a Dockerfile that describes our desired environment
- We have built a docker image that contains all
- We have a default command so a user can i
- -> **We have containerized the classifier**

The bind-mount will be created in the container, even if it exists already!

What if I'd like to make a change to a

```
$ docker build  
$ docker build  
$ docker build  
$ docker build  
$ docker build  
$ docker build .
```



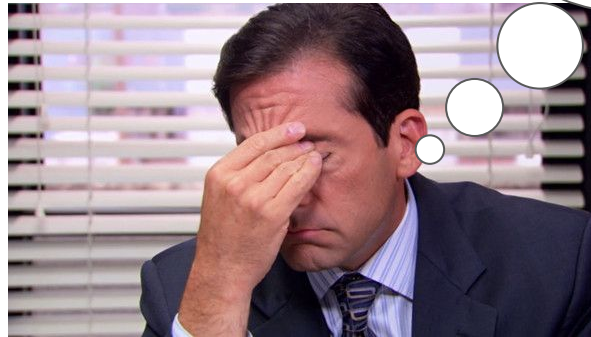
Exercise 3: Containerize a classifier

- We have a Dockerfile that describes our desired environment
- We have built a docker image that contains all the dependencies
- We have a default command so a user can just run the container
- -> **We have containerized the classifier**

The bind-mount will be created in the container, even if it exists already!

What if I'd like to make a change to the classifier?

```
$ docker build  
$ docker build  
$ docker build  
$ docker build  
$ docker build  
$ docker build .
```



Exercise 3: Keep developing the containerized classifier

- I can bind-mount my local source code over the copied code in the container
- I can keep editing my files locally and test them in the container
- When I build a new image, the most recent files will be copied into the image



Exercise 3: Containerize a classifier

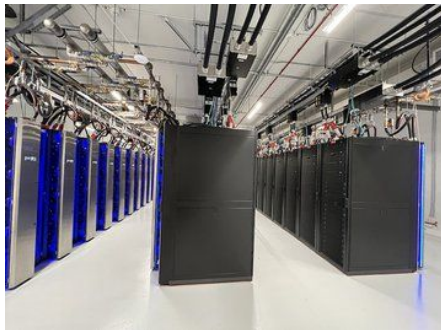
Summary:

- Dockerfile + docker build create a **controlled image** for our code
- bind-mounts allow us to **expose data** to the container and store results
- We can combine **local editing** of source files and the **controlled environment** of the container by bind-mounting our source code
- -> We can use the **same environment** for development, analysis, and sharing

Exercise 3: Containerize a classifier

Summary:

- Dockerfile + docker build create a **controlled image** for our code
- bind-mounts allow us to **expose data** to the container and store results
- We can combine **local editing** of source files and the **controlled environment** of the container by bind-mounting our source code
- -> We can use the **same environment** for development, analysis, and sharing



Can I use my new container on a supercomputer?

Singularity

Singularity is the container solution for HPCs

On shared systems (like a supercomputer), you shouldn't / can't use Docker

- Docker isn't as isolated as a VM
- By default you run docker with root privileges and are root inside a container
- A malicious actor can escalate privileges and “break out” of a container

Apptainer (formerly Singularity) is a container solution in these cases

Singularity^[1] is open source software created by Berkeley Lab:

- as a **secure way** to use Linux containers on Linux multi-user clusters,
- as a way to enable users to have **full control of their environment**, and,
- as a way to **package scientific software** and deploy such to *different* clusters having the *same* architecture.

i.e., it provides **operating-system-level virtualization** commonly called *containers*.

Build images with Docker, run them with Singularity

1. Pull an image from Dockerhub and create a local **SingularityImageFile**

```
$ apptainer pull docker://sylabsio/lolcow
```

2. Run the Singularity File using **apptainer run**

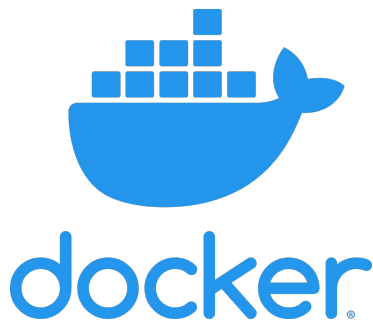
```
$ apptainer run lolcow_latest.sif
```

```
< Mon Aug 16 13:01:55 CDT 2021 >
```

```
-----  
  \  ^ ^  
   (oo)\_____  
    (__)\       )\/\  
       ||----w |  
       ||     ||
```

Container Summary

- Docker **images** are **read-only snapshots**, you can find them on Dockerhub
- Images have tags (or version), that you should specify when pulling
- A Docker **container** is an isolated, **live instance** of an image
- Docker **containers** have their own file system, but this is **not persistent**
- With **volumes** we can **expose directories** on the host to the container
- We can **build** our own images on top of existing ones using a **Dockerfile**
- Use **Singularity/Apptainer** to run Docker images on a compute cluster



Docker engine

```
1 # our base image
2 FROM alpine:3.5
3
4 # Install python and pip
5 RUN apk add --update py2-pip
6
7 # upgrade pip
8 RUN pip install --upgrade pip
9
10 # install Python modules needed by the Python app
11 COPY requirements.txt /usr/src/app/
```

Dockerfile



Docker image registry

There are tools to help make Dockerfiles

Welcome to Neurodocker!

Neurodocker is a command-line program that generates custom Dockerfiles and Singularity recipes for neuroimaging and minifies existing containers. Its purpose is to make it easier for scientists (and others) to easily create reproducible computational environments.

(This requires having [Docker](#) installed)

```
neurodocker generate docker --pkg-manager apt \  
  --base-image neurodebian:buster \  
  --ants version=2.3.4 \  
  --miniconda version=latest conda_install="nipy notebook" \  
  --user nonroot
```

Additional Resources

- Docker tutorial <https://github.com/docker/labs>
- Neurohackweek container course
<https://neurohackweek.github.io/docker-for-scientists/>
- The Turing Way on reproducible research environments
<https://the-turing-way.netlify.app/reproducible-research/renv.html>