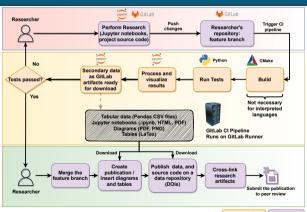
"Continuous" Integration of Scientific Software (in Computational Science and Engineering)



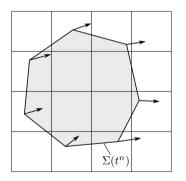
Students@SC21 2021-11-17

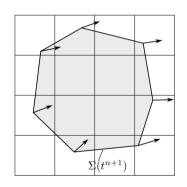


Motivation: multiphase flow simulation methods

Lagrangian / Eulerian Interface Advection (LEIA) methods







- Fluids that do not mix are separated by an interface $\Sigma(t)$ (surface in 3D).
- Goal: track $\Sigma(t)$ as it moves in time t and changes its topology.

Motivation: multiphase flow simulation software

Lagrangian / Eulerian Interface Advection (LEIA) Methods



LEIA methods ^{1, 2, 3, 4, 5} require **thorough testing**:

- **Verification** cases: evolution of $\Sigma(t)$ and two-phase flows with exact solutions.
- Validation with respect to experiments.
- Testing serial and parallel computational efficiency.

¹Marić, T., Marschall, H., & Bothe, D. (2015). lentFoam–A hybrid Level Set/Front Tracking method on unstructured meshes. Computers & Fluids, 113, 20-31.

²Tolle, T., Bothe, D., & Marić, T. (2020). SAAMPLE: A Segregated Accuracy-driven Algorithm for Multiphase Pressure-Linked Equations. Computers & Fluids, 200, 104450.

³Marić, T., Kothe, D. B., & Bothe, D. (2020). Unstructured un-split geometrical Volume-of-Fluid methods – A review. Journal of Computational Physics, 420, 109695.

⁴Marić, T. (2021). Iterative Volume-of-Fluid interface positioning in general polyhedrons with Consecutive Cubic Spline interpolation. Journal of Computational Physics: X, 11, 100093.

⁵Tolle, T., Gründing, D., Bothe, D., & Marić, T. (2021). Computing volume fractions and signed distances from triangulated surfaces immersed in unstructured meshes. arXiv preprint arXiv:2101.08511.

Computational Science and Engineering software in university research groups Boundary and initial conditions



- Publish or perish prioritizes publications over scientific software.
- Dedicated resources for increasing software quality are usually not available.
- Ph.D. students rotate every 3-5 years, postdocs every 1-2 years.
 - Little or no overlap between successors and predecessors.
- Large-scale software design is not a mandatory part of the CSE curriculum.
 - Different CSE background: (Applied) Mathematics, Mechanical Engineering, Physics, Informatics.

⁶Symbol of a publish-or-perish simplification of the workflow :)

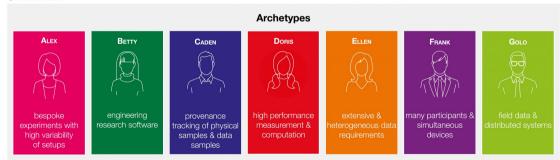
NFDI4Ing to the rescue!

Resources for engineering research software



NFDi4ing

productive since 2017



NFDI4Ing resources.

BSSW to the rescue!

Resources for engineering research software





Better Scientific Software resources.

Computational Science and Engineering software in university research groups The chaos scientific legacy code





Betty is a CSE researcher, working with a legacy research code. Why is Betty so (rightfully) angry?

- Betty inherited a research software that is only partially tested.
- Betty inherited a research software that isn't automatically tested.
 - Betty changes one part of the code and gets her model running, only to see 10 other things fail, after days of manually running tests.
- Betty's software has no documentation of the scientific workflow.
 - Betty doesn't know how to use existing scripts to run simulations and analyze (reproduce) results.
- Betty's software has disjoint (diverging) versions that she can't integrate.
- Betty can't even find code versions used to generate results in the publications from her research group.

Computational Science and Engineering software in university research groups The chaos of developing entirely new research software



"Après moi, le déluge" - "After me, the flood"

Louis XV of France

Research software generally does not matter, as long as papers are published ().

Missed opportunities

- Finding results made easy by cross-linking code versions, data and publications.
- Faster extension / combination of existing ideas if their respective versions are integrated.
- Faster comparison of results with previous ideas automating verification / validation.
- Automatic reproducibility of results using automated testing and version control.
- Faster onboarding with documented scientific verification and validation workflows.

Computational Science and Engineering software in university research groups



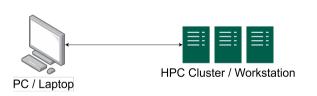
Continuous integration and cross-linking to the rescue

Automated testing (verification and validation), **version control**, and **cross-linking** reports, source code and research data increase Findability, Accessibility and Reproducibility (FAIR) and **speed up research**.

- Continuous Integration (CI) = automatic testing + version control.
- CSE research requires scientific workflows: initialize simulations, run parameter variations, agglomerate data, visualize, and check results.
- CI can be used to automate and document scientific workflows.
- CI ensures that the integration of new changes does not break existing functionality.
- Once the changes are integrated, the publication, the source code and the data are published on pre-print and data repositories and cross-linked using git tags and DOIs.

Research Software Workflow I





while Results are unsatisfactory do

Work on algorithms.

(Compile the code.)

for All studies do

Prepare the study.

Run the study.

Analyze results.

Move results to a report.

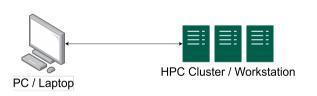
end for

Compare old and new results.

end while

Research Software Workflow II



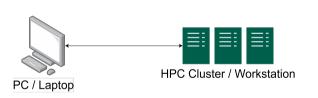


Issues...

- Starting studies takes time.
- Analyzing results takes time.
- Often the results are not checked "live" as the study runs - waste of research time and CPUh.
- Only the researcher knows the details behind the initialization, running and post-processing scripts - when this person leaves, the reproducibility is gone.
- A researcher may forget to run a study and believe all tests have passed.

Automating the research workflow I





while Results are unsatisfactory do

Work on algorithms.

(Compile the code.)

for All studies do

Prepare the study.

Run the study.

Analyze results.

Move results to a report.

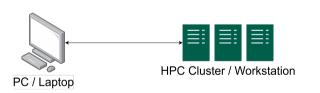
end for

Compare old and new results.

end while

Automating the research workflow I





while Results are unsatisfactory do
Work on algorithms.
(Compile the code.)
Run initialization scripts (jobs).
Run simulation scripts (jobs).
(Run postprocessing scripts (jobs)).
Visualize results live in Jupyter notebooks.

end while

Automating the research workflow II



Manual steps of the research workflow,

(Compile the code.)

for All studies do

Prepare the study.

Run the study.

Analyze results.

Manage results.

Move results to a report.

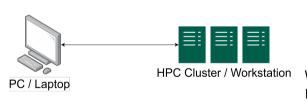
end for

Compare old and new results.

are now automated using scripts that do not require additional knowledge / input (metadata).

Automating the research workflow III





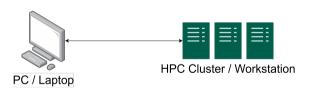
- 1. The **new** results are satisfactory.
- Similar automated workflows are executed for existing tests.
- 3. All results are checked.
- 4. The milestone has been reached, the version can be integrated.

Works well manually when there aren't many previous verification/validation tests and their analysis is relatively simple.

Are we sure that we ran all the tests and examined the results properly?

Automating the research workflow III

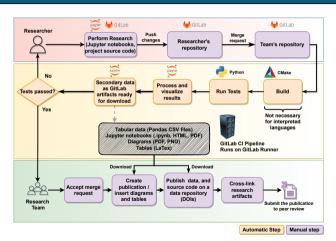




- Manual testing takes a lot of time.
- Manual testing of all previous tests is prone to error - even if V&V scripts do not require metadata.
- Relevant V&V tests are automated using Continuous Integration (CI).
 - Changes are pushed to the upstream version control repository.
 - The remote repository starts the so-called
 CI test pipeline (a sequence of tests).
 - Tests are automatically run, processed and visualized.

Schematic diagram for the team workflow





Working in a team.



- A text (YAML) file is added to a repository, that specifies the tests (jobs) in a CI pipeline.
- When the YAML file is pushed to an upstream git repository (GitLab), GitLab creates a CI pipeline from the YAML file.
- The CI pipeline needs a machine for running tests the GitLab runner.
 - Shared runners on gitlab.com have limited capacity.
 - We can install and register our own GitLab runner.
- A Docker image encapsulates the computing environment.
 - Virtualization/Containerisation increases reproducibility and simplifies testing.
- The Docker image must be publicly accessible for it to be used by a shared runner.



```
initialization_param_study:
    stage: running
    dependencies:
        - build_release
    script:
        # run the parameter variation tests
        - cd cases/initialization/3dinit
        - ./create_and_run_levelset.sh
        - ./reproduce_publication_results.sh
    artifacts:
        paths:
```

- cases/initialization/3dinit/*.csv

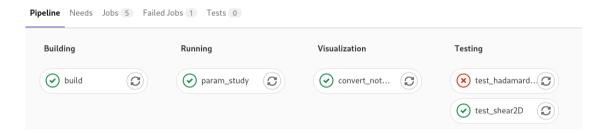
- cases/initialization/3dinit/*.pdf

Example YAML file

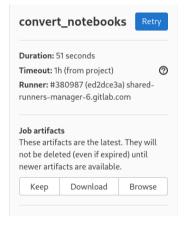
- The CI pipeline starts the right scripts in the right order: it documents the research workflow.
- A click of a button in a web browser reproduces results for any version of the research software.
- Continuous integration is used to integrate only those changes that improve the software and don't break existing tests.



An example CI pipeline



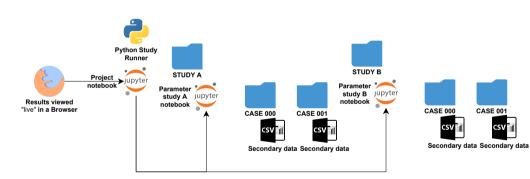




- Files created within a CI job are gone when the job ends.
- GitLab uses job artifacts to pass on data from one job to the next.
- Job artifacts can only be files stored in project's sub-folders.
- Libraries and applications are passed to other jobs as artifacts.
- Artifacts can be downloaded on the GitLab project website.

(Continuous) Integration of scientific software Running tests I





Organize your simulation studies.

(Continuous) Integration of scientific software Running tests II



- Success of CSE methods is measured using verification and validation data.
- Effective comparison with others (previous versions) hinges on data organization.
- Goal: easily programmatically identify parameters used in a simulation case.

Legacy code:

- use the existing folder structure and parameterization tools
- $lue{}$ The mapping (case000) ightarrow (parameter vector) must be stored (YAML, ...)

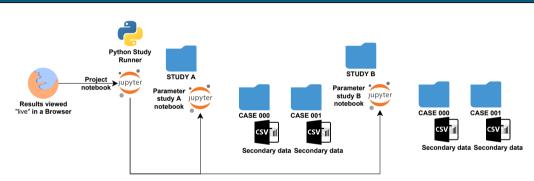
New code:

- 1. Simple folder and file structure
- 2. HDF5⁷ or other open data format.
- 3. Alternative to HDF5: ExDir8

⁷https://www.hdfgroup.org/solutions/hdf5

⁸Dragly, Svenn-Arne, et al. "Experimental Directory Structure (Exdir): An alternative to HDF5 without introducing a new file format." Frontiers in neuroinformatics 12 (2018): 16.





- Associate simulation cases with their metadata.
- {case000 : {N_CELLS: 32, MODEL : shear2D}}
- Store this information using a standard open-source format (Interoperability in FAIR).



Use Jupyter notebooks⁹ and pandas¹⁰ for

- **Documentation**: geometry, initial and boundary conditions, error norms, comparison data.
- Data processing: verification errors (conservation, convergence, stability), validation errors
- Result analysis: interactive and remote, while simulations are running!

T. Marić, T. Tolle, JP. Lehr, I. Pappagianidis, B. Lambie, D. Bothe, C. Bischof

⁹https://jupyter.org/

¹⁰ https://pandas.pydata.org/

Processing and visualizing tests



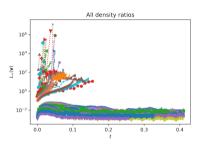
jupyter nbconvert notebook.ipynb --execute --to FORMAT

- Agglomerate secondary data into pandas.MultiIndex CSV files.
- Run each jupyter notebook in the repository.
- Export secondary data and notebooks in different formats as artifacts.
- Visualization
 - Download the artifact and open the notebook <a>.
 - Notebooks contain information on failing tests.
 - $lue{}$ Mapping "caseID" ightarrow "parameters" is crucial for re-starting failed parameter variations!

(Continuous) Integration of scientific software Secondary data I



- Data used for diagrams and tables in a publication.
- Data we compare our results with.
- Data we waste time scanning from (sometimes our own) publications in CSE.



(a) Old inconsistent method: interface stable only for cases with density ratio $\rho^-/\rho^+=1$

Imagine scanning this diagram.
Preprint: https://arxiv.org/abs/2109.01595
Data: https://doi.org/10.48328/tudatalib-627



pandas. MultiIndex CSV with metadata for secondary data

- pandas.MultiIndex saved in "metadata columns".
- Metadata is repeated: not an issue for the small secondary data!
- lacktriangledown Metadata in columns o pandas.MultiIndex o strongly simplified data analysis.
- Direct readable export of tables to LaTeX!

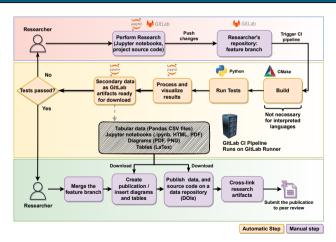
	Н	L_INF	O(L_INF)	EPSILON_R_EXACT_MAX	O(EPSILON_R_EXACT_MAX)
VELOCITY_MODEL					
SHEAR_2D	0.125000	0.032961	1.833407	0.032961	1.833407
SHEAR_2D	0.062500	0.009249	1.955529	0.009249	1.955529
SHEAR_2D	0.031250	0.002385	1.988745	0.002385	1.988745
SHEAR_2D	0.015625	0.000601	1.997178	0.000601	1.997178
SHEAR_2D	0.007813	0.000150	1.999294	0.000150	1.999294
SHEAR_2D	0.003906	0.000038	1.999294	0.000038	1.999294

[&]quot;Continuous" Integration of Scientific Software(in Computational Science and Engineering) -

T. Marić, T. Tolle, JP. Lehr, I. Pappagianidis, B. Lambie, D. Bothe, C. Bischof

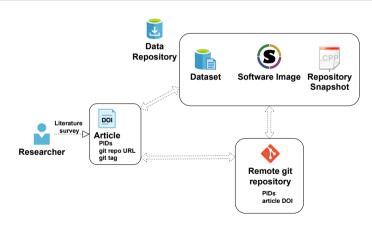
Schematic diagram for the individual workflow





Working alone.

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Cross-linking I



Cross-linking is done manually.

Cross-linking II

- Place whatever you can under version control.
- When a set of milestones is reached (release), use git-tags as version snapshots, and upload the research data to a data repository, e.g. TUDatalib at TU Darmstadt, or Zenodo.
 - Secondary data (diagrams, tables), raw data (simulations, experiments), archive of the research software, ...
- Data uploaded to a data repository is associated with Persistent Identifiers (PIDs), e.g. DOIs.
- Cite the research data using DOIs in the report (article, preprint).
- Upload the report to a pre-print repository, e.g. ArXiv.
- Edit the data on the data repository and mention the arXivID.
- Submit the pre-print to a journal for peer-review.

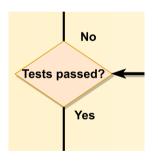
Cross-linking III



- Research software is compared with existing publications.
- A major milestone are improved results for a set of verification / validation tests.
- The cross-linking therefore revolves around the publication (pre-print, report, ...).
- The cross-linking makes it possible to find the version of research software used to generate the results in the publication: repository link + git tag, repository snapshot, software image.
- Once the version is found, CI automatically reproduces all results from the publication with a click of a button.

(Continuous Integration with result visualization)





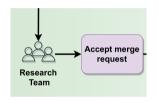
Test evaluation I

Straightforward for easily quantifiable errors

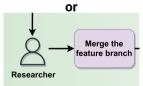
- Examples: volume conservation, order of convergence, total wall clock time, weak scaling, ...
- Python scripts test secondary data agglomerated by Jupyter notebooks from simulation results.

(Continuous Integration with result visualization)





Test evaluation II

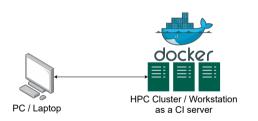


Difficult for errors that cannot be quantified easily

- Examples:
 - Is is the difference between simulation and experiment data ≤ 4%?
 - How to quantify the difference for complex signals?
- Option 1: Researchers evaluate the test results even if all CI jobs pass.
 - A simple and efficient solution
- Option 2: Use statistics to quantify the difference.

Docker (containerization)

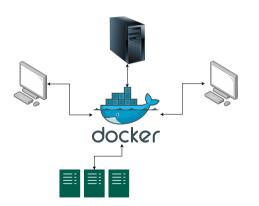




- Instead of installing the research software only on the laptop/PC and the HPC cluster / workstation, we install it in a virtual environment - a Docker image.
- The Docker image then works on any machine that runs Docker.
- Sharing research software becomes trivial if our colleague wants to use our software, no installation (besides Docker) is required.

Docker (containerization)





- Instead of installing the research software only on the laptop/PC and the HPC cluster / workstation, we install it in a virtual environment - a Docker image.
- The Docker image then works on any machine that runs Docker.
- Sharing research software becomes trivial if our colleague wants to use our software, no installation (besides Docker) is required.

Computing resources



The GitLab CI requires a **GitLab runner**: a machine that runs the **CI jobs**.

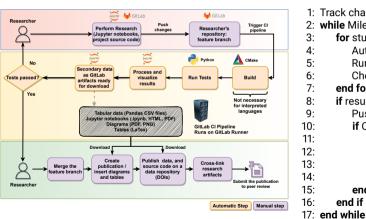
- 1. Short few CPU-core tests: work-PC 🎓.
- 2. Short many-core tests: obtain a workstation with a 64-Core CPU¹¹ .
- 3. **HPC tests**: combine 1. or 2. with an HPC cluster.

An HPC cluster is relevant for production tests and performance measurements.

- This workflow uses coarse ("smoke") tests 🕿
 - Unit tests run for 1. and 2.
 - Convergence ensured for 1. and 2.
 - Is efficient in parallel for 1. and 2.
- Challenge: Is it possible to combine 1., 2. and 3. and publish instead of perish ?

¹¹Thanks to CRC 1194 at TU Darmstadt.





```
1: Track changes using version-control.
 2: while Milestone not reached do
3:
       for study in studies do
                                       ⊳ On an HPC cluster.
4:
          Automate data processing and visualization.
          Run study.
6:
          Check results and apply code changes.
       end for
8:
       if results are improved on the HPC cluster then
g.
          Push changes to the remote repository.
10:
           if CI pipeline tests pass then
11:
              Milestone reached
12.
              Add new tests to the CI pipeline.
13:
              Merge feature into development branch.
14.
              Cross-link publication, data, and source code.
15:
          end if
16:
       end if
```

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Similarity with other workflows / best practices

Our (subjective) estimates* of similarity 1-5 (higher is more similar), -: aspect not addressed.

DOI	Branching model	TDD	Cross-linking	CI	(Meta)data standardization
10.12688/f1000research.11407.1	-	-	-	-	1
10.3934/math.2016.3.261	-	-	-	-	2
10.1371/journal.pbio.1001745	1	2	-	-	-
10.1371/journal.pcbi.1005510	-	-	3	1	3
10.1145/2723872.2723881	1	-	-	1	-
10.1145/3324989.3325719	1	-	-	5	-
10.1371/journal.pone.0230557	1	-	-	1	4
10.1145/3219104.3219147	1	-	-	4	-

^{*}The list may still be incomplete.

Hands on: overview



1. Repository preparation

A minimal repository representing an exemplary "status quo".

2. Create a Docker image

Configure a reproducible testing environment.

3. Define CI pipeline through a YAML file

Define tests, how and when they are executed and what results to store.

4. Setup your own GitLab runner

Provide a machine for execution of tests.

Hands on: prepare the example repository



Create your own copy of the example repository by forking:

- Log in to https://gitlab.com/.
- Go to https://gitlab.com/tmaric/minimal-cse-ci-examples.
- Click fork (upper right corner).
- Select a namespace, e.g. your personal one.
- Select either Private or Public as visibility level, both are fine.
- Click Fork project.
- Clone your fork on your machine:
 - ?> git clone your-fork-URL

Hands on: overview



1. Repository preparation

2. Create a Docker image

A minimal repository representing an exemplary "status quo".

Configure a reproducible testing environment.

3. Define CI pipeline through a YAML file

Define tests, how and when they are executed and what results to store.

4. Setup your own GitLab runner

Provide a machine for execution of tests.

Hands on: install Docker I



Specific steps depend on your Linux distribution (Docker documentation)
Here for Ubuntu Focal:

- 1. ?> sudo apt-get update
- 2. ?> sudo apt-get install apt-transport-https ca-certificates curl gnupg lsb-release
- 4. ?> echo \
 - "deb [arch=amd64 signed-by=/usr/share/keyrings/docker-archive-keyring.gpg] \
 https://download.docker.com/linux/ubuntu \
 \$(lsb_release -cs) stable" | sudo tee /etc/apt/sources.list.d/docker.list > /o
 - \$(lsb_release -cs) stable" | sudo tee /etc/apt/sources.list.d/docker.list > /dev/null
- 5. ?> sudo apt-get update
- 6. ?> sudo apt-get install docker-ce docker-ce-cli containerd.io

Hands on: install Docker II



Check your Docker installation by running

?> sudo docker run hello-world The output should look as shown on the right.

Hands on: creating a Docker image



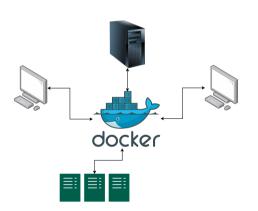
In the minimal-cse-ci-examples repository

?> git checkout starting-point

?> git checkout -b feature/dockerfile

Hands on: creating a Docker image I





- Docker images are computing environments that contain (dependencies) needed to build the research software, run simulations and process results.
- Sharing docker images removes the need to install the dependencies on different machines.
- The computing environment in a Docker image is usually based on an existing Linux distribution.
- The Docker image is built from a text file, that specifies installation steps for the dependencies, the so-called **Dockerfile**.

Hands on: creating a Docker image II



In a file named 'minimal-cse-ci-dockerfile_ubuntu-focal', write

FROM ubuntu:focal

```
# Set timezone
RUN apt-get update --fix-missing && \
    DEBIAN_FRONTEND="noninteractive" apt-get -y install tzdata
```

- We'll use Ubuntu 20.04 (focal) as the base system.
- Steps that are usually done manually (setting the timezone) are automated.

Hands on: creating a Docker image III



Dependency installation

```
# Install packages
RUN apt update && apt-get install --fix-missing -v \
    # Building
    build-essential cmake \
    # Version control
    ait \
    # Python
    pvthon3 \
    # Visualization
    python3-matplotlib python3-numpy \
    # Data analysis
    python3-pandas \
    # Test visualization
    jupyter-notebook jupyter-nbconvert \
    # Debugging the image
    vim
```

- RUN runs commands in the Docker container.
- The Docker container is a process spawned using the Docker image as the computing environment.
- Install the software needed for the scientific workflow (dependencies).

Hands on: creating a Docker image IV



Software setup

Default Ubuntu to python3

RUN update-alternatives --install \
 /usr/bin/python python /usr/bin/python3 10

Some specifics

- Alternative (g++) compiler.
- Alternative working directory.

Hands on: creating a Docker image V



Complete Dockerfile for the minimal example

```
FROM ubunturfocal
# Set timezone
RUN apt-get update --fix-missing && \
   DEBIAN FRONTEND="noninteractive" apt-get -v install tzdata
# Install packages
RUN ant undate && ant-get install --fix-missing -v \
    # Building
   build-essential cmake \
   # Version control
   ait \
   # Python
   nython3 \
    # Visualization
   python3-matplotlib python3-numpy
   # Data analysis
   python3-pandas \
    # Test visualization
    iupyter-notebook iupyter-nbconvert \
   # Debugging the image
    vim
## Default Ubuntu to python3
RUN update-alternatives --install \
```

- The example Dockerfile installs all dependencies for the minimal example on Ubuntu 20.04
- The installation commands would be different for another operating system.
- A more complex software (e.g. OpenFOAM) requires a larger Dockerfile.
- This lets us define the computing environments that are supported by the research software.

/usr/bin/pvthon pvthon /usr/bin/pvthon3 10

Hands on: creating a Docker image VI



Building the image

```
%?> sudo docker build . \
    %-f minimal-cse-ci-dockerfile_ubuntu-focal \
    %-t minimal-cse-ci-dockerfile_ubuntu-focal
    sudo docker build . -f minimal-cse-ci-dockerfile_ubuntu-focal -t mi
```

- ".": current directory
- "-f" name of the Dockerfile (defaults to "Dockerfile")
- "-t" tag (name) of the Docker image

Hands on: creating a Docker image VII



Listing Docker images

?> sudo docker image list
REPOSITORY TAG IMAGE ID CREATED SIZE
minimal-cse-ci-dockerfile_ubuntu-focal latest 921233ec4b44 9 minutes ago 982MB

- The image is built on the machine (host) where the docker build command is called.
- Docker uses a so-called image registry to store images.
- For Continuous Integration the images are built on the machine where the tests are run or shared on Dockerhub.

Hands on: creating a Docker image VIII



"Spinning a container" (running a Docker image)

```
?> sudo docker run -it minimal-cse-ci-dockerfile ubuntu-focal /bin/bash
root@b2c14ee0fd58:/# ls
    boot dev etc home lib
                               1ib32
                                      1ib64
                                             libx32
                                                     media
hin
                                                            mnt
    proc root run sbin srv
opt
                                svs
                                     tmp
                                          usr
root@b2c14ee0fd58:/# cd
root@b2c14ee0fd58:~# pwd
/root
```

- The container behaves just like a "regular" Ubuntu.
- Jobs (test) commands for the Continuous Integration are checked/debugged inside a running container.
 - Forgot to install a dependency.
 - The research software does not compile with installed dependencies.
 - **-** ...

Hands on: creating a Docker image IX



Working within the container: compiling the software

- ?> git clone https://gitlab.com/tmaric/minimal-cse-ci-examples.git
- ?> cd minimal-cse-ci-examples && mkdir build && cd build
- ?> cmake .. && make
- ?> ./myapp

The same steps will be done in the Docker container by the Continuous Integration

- Clone the repo.
- Build the software.
- Run the tests.

Hands on: creating a Docker image X



Analyzing the data using Jupyter notebooks

```
?> cd ..
```

?> jupyter nbconvert --execute mynotebook.ipynb --to html

- On the cluster, one would start the Jupyter notebook server and connect to it locally.
- Here the notebook is used to process the results and visualize secondary data as tables and diagrams.

Hands on: creating a Docker image XI



Extracting the data from the container:

- Find the ID of the container you're on (execute on your machine) sudo docker ps
- Copy the results from the container onto the local machine (execute on your machine) mkdir container-data sudo docker cp f2dff55edf7a:/root/minimal-cse-ci-examples \ container-data/
- Examine the data and the Jupyter notebook in a browser.

Note: the sequence f2dff55edf7a is system dependent ID, so it'll be different for you.

Hands on: creating a Docker image XII



- Saving the container or an image as a tar file sudo docker commit f2dff55edf7a test:latest
- You can exit/close the container by pressing Ctrl+d.
- View the newly create image with 'name:tag' using sudo docker image list

REPOSITORY TAG IMAGE ID CREATED SIZE test latest f2dff55edf7a About a minute ago 983MB

- Save the image into a tar file sudo docker save test:latest -o container-archive.tar
- Load an image into Docker's registry to work with it sudo docker load -i container-archive.tar

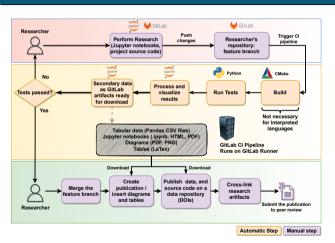
Hands on: creating a Docker image XIII



- Usually, the Docker image "lives" locally on the test machine.
- However, it can also be shared publicly on Dockerhub, for example (don't do this now)
 - ?> docker login
 - ?> docker tag name:tag username/name:tag
 - ?> docker push username/name:tag
- This image can now be used by everyone.
- Note: once you exit/stop a container all data/files created inside the container are discarded.

Hands on: creating a Docker image XIV





All the steps done so far manually using Docker, namely,

- 1. **building** the scientific software,
- 2. running tests,
- 3. processing data
- exporting the data and Jupyter notebooks,

are automated by Continuous Integration, that uses Docker for encapsulating the computing environment.

Hands on: overview



1. Repository preparation

A minimal repository representing an exemplary "status quo".

2. Create a Docker image

Configure a reproducible testing environment.

3. Define CI pipeline through a YAML file

Define tests, how and when they are executed and what results to store.

4. Setup your own GitLab runner

Provide a machine for execution of tests.

Hands on: enabling CI for a GitLab project



In the **minimal-cse-ci-examples** repository

- ?> git checkout added-dockerfile
- ?> git checkout -b feature/enable-ci

Hands on: enabling CI for a GitLab project I



- Adding the .gitlab-ci.yml file your project and pushing the change to the GitLab remote repo configures the CI pipeline.
- The YAML file specifies the Docker image that is used for testing image: "tmaric/minimal-cse-ci:ubuntu-focal"

stages:

- building
- running
- visualization
- and the so-called job **stages**: collections of jobs for building, running tests and visualization.
- For example, the building stage may multiple jobs, building the software for
 - production,
 - debugging,
 - performance measurements.

Hands on: enabling CI for a GitLab project II



The building stage in the YAML file defines build jobs like this one build:

```
stage: building
script:
    - git clone https://gitlab.com/tmaric/minimal-cse-ci-examples.git
    - cd minimal-cse-ci-examples && mkdir build && cd build
    - cmake ...
    - make
artifacts:
  paths:
      - minimal-cse-ci-examples/mynotebook.ipynb
```

- minimal-cse-ci-examples/build/myapp
- where the repository is cloned and built with specific options.
- For example cmake -DCMAKE_BUILD_TYPE=Debug can set up the build for debugging.
- **artifacts** are downloadable files passed on to other jobs.

Hands on: enabling CI for a GitLab project III



The running stage in the YAML file defines how simulations (studies) run param_study:

```
stage: running
dependencies:
   - build
```

script:

- cd minimal-cse-ci-examples/build && ./myapp

artifacts: paths:

- minimal-cse-ci-examples/mynotebook.ipynb
- minimal-cse-ci-examples/build/myapp
- minimal-cse-ci-examples/build/poly-data.csv
- Without a successful build, simulations do not run.
- Here the artifacts are the secondary data and the notebooks that visualize them.

Hands on: enabling CI for a GitLab project IV



■ The **visualization** stage in the YAML file saves time by converting Jupyter notebooks

```
convert_notebooks:
    stage: visualization
    dependencies:
        - param_study

script:
        - cd minimal-cse-ci-examples
        - jupyter nbconvert mynotebook.ipynb --execute --to html

artifacts:
    paths:
        - minimal-cse-ci-examples/mynotebook.*
        - minimal-cse-ci-examples/build/myapp
```

- minimal-cse-ci-examples/build/polydata.csv
- HTML is easiest, other formats are available (PDF, markdown,...).
- HTML notebooks can be viewed in the browser.

Hands on: enabling CI for a GitLab project V



Lessons learned I

- Defining artifacts path starts at your-project/.
- YAML files require debugging:
 - syntax: use GitLab's CI Lint tool
 - everything else: the only way to do this effectively is to commit changes and push them upstream.
- It is possible to partially debug locally using gitlab-runner exec docker job-name but this does not work with artifacts and dependencies.

Final **.gitlab-ci.yml** file.

Hands on: enabling CI for a GitLab project VI



Final .gitlab-ci.yml file.

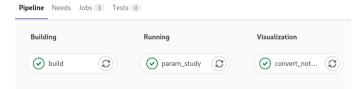
Lessons learned II

- Generally, and for the CI, scripts that reproduce data without requiring input for the users speed up work. simulation-directory > ./reproduce-density-ratio-data
- It takes time to set up the CI, but it pays off in debugging time as problems are found automatically.
- Exporting *.ipynb jupyter notebooks and their data

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Hands on: enabling CI for a GitLab project VII

The CI pipeline of the Minimal Working Example (MWE) repository



Hands on: overview



1. Repository preparation

A minimal repository representing an exemplary "status quo".

3. Define CI pipeline through a YAML file

Define tests, how and when they are executed and what results to store.

2. Create a Docker image

Configure a reproducible testing environment.

4. Setup your own GitLab runner

Provide a machine for execution of tests.

Hands on: setup a GitLab runner I (why a self-managed runner)



An incomplete comparison:

Self-managed runner

- No shared runners available
- Provided CI/CD minutes of plan insufficient (e.g. 400 min per month for GitLab's free plan)
- Control over hardware and runner configuration

Shared runners

- No need to provide "always on" hardware
- No need for maintenance

Overall: require less of your time

Hands on: setup a GitLab runner II (install GitLab runner)



Using instructions from GitLab documentation for Ubuntu:

- 2. ?> sudo apt-get install gitlab-runner

Check the status of the runner:

• ?> sudo systemctl status gitlab-runner.service

The output should indicate that it is active:

Hands on: setup a GitLab runner III (register GitLab runner)



Follow GitLab documentation on how to register a runner:

- Obtain a token for project-specific runner: go to your fork of the minimal-cse-ci-examples on gitlab.com and then to Settings > CI/CD and expand the Runners sections.
- There you find a section Specific runners and aforementioned token.

Register your runner (instructions for Linux):

?> sudo gitlab-runner register

You need to provide some information regarding your runner, e.g. your project's token. See next slide.

Hands on: setup a GitLab runner IV (register GitLab runner)



Option	Value
GitLab instance URL	https://gitlab.com/
Token	Obtained from the project on GitLab, see previous slide
Runner description	describe the machine used as runner, useful to distinguish multiple runners
Tags	leave empty, not required here. Useful for advanced pipelines
Runner executor	docker (see here for comparison of executors.)
Default image	Because we chose docker as executor: name of the default Docker image

You should now see your runner under *Available specific runners*:

Available specific runners





Acknowledgements





Funded by the German Research Foundation (DFG) - Project-ID 265191195 - CRC 1194 : Z-INF