

Good software engineering practices

Oliver Beckstein

Summer 2018 Beckstein Lab Mini-Workshops
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<https://becksteinlab.physics.asu.edu/learning/117/summer-2018-mini-workshops>



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Role of Software in Academic Research

science

Theory

Experiment

Computation/
Simulation

nature

What **fraction of your projects** requires scientific software?

1. $< 10\%$
2. $10\% - 30\%$
3. $30\% - 50\%$
4. $> 50\%$

How much of your time do you spend on **writing software**?

1. < 10%
2. 10% – 30%
3. 30% – 50%
4. > 50%

What do you consider **most important** in scientific software that you use?

1. fast
2. graphical user interface
3. correct
4. has documentation
5. source available
6. user-friendly
7. written in specific programming language (FORTRAN, C, C++, Python, rust, Ruby, Julia, ook, assembly, ...)
8. open source (licence e.g., BSD-3, MIT, GPL, CC0, ...)
9. support available
10. free (\$\$\$)
11. runs on specific OS (Linux, macOS, Windows) OR architecture (x86, GPU, Phi, ...)

...why scientific programming does not compute



...SCIENTISTS AND THEIR SOFTWARE

A survey of nearly 2,000 researchers showed how coding has become an important part of the research toolkit, but it also revealed some potential problems.

> **45%** said scientists spend more time today developing software than five years ago."

> **38%** of scientists spend at least one fifth of their time developing software.

> Only **47%** of scientists have a good understanding of software testing.

> Only **34%** of scientists think that formal training in developing software is important.

...PRACTICING SAFE SOFTWARE

> Five tips to make scientific code more robust.

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Put source code, raw data files, parameters and other primary material into it to record what you did, and when.

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Build large codes from smaller, easily testable chunks.

← Test the software:

And get somebody else to read it and look for bugs.

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Make the code that you use in research freely available, when possible.

... ERROR

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Ten Simple Rules for Reproducible Computational Research

Geir Kjetil Sandve^{1,2*}, Anton Nekrutenko³, James Taylor⁴, Eivind Hovig^{1,5,6}

Citation: Sandve GK, Nekrutenko A, Taylor J, Hovig E (2013) Ten Simple Rules for Reproducible Computational Research. PLoS Comput Biol 9(10): e1003285. doi:10.1371/journal.pcbi.1003285

1. For every result, keep track how it was produced
2. Avoid manual data manipulation steps
3. Archive the exact versions of all external programs used
4. Version control all scripts
5. Record all intermediate results, when possible in standardized formats
6. For analyses that include randomness, note underlying random seeds
7. Always store raw data behind plots
8. Generate hierarchical analysis output, allowing layers of increasing details to be inspected
9. Connect textual statements to the underlying result
10. Provide public access to scripts, runs, and results

Ten Simple Rules for Effective Computational Research

James M. Osborne^{1,2*}, Miguel O. Bernabeu^{3,4}, Maria Bruna^{1,2}, Ben Calderhead³, Jonathan Cooper¹, Neil Dalchau², Sara-Jane Dunn², Alexander G. Fletcher⁵, Robin Freeman^{2,3}, Derek Groen⁴, Bernhard Knapp⁶, Greg J. McInerny^{1,2}, Gary R. Mirams¹, Joe Pitt-Francis¹, Biswa Sengupta⁷, David W. Wright^{3,4}, Christian A. Yates⁵, David J. Gavaghan¹, Stephen Emmott², Charlotte Deane⁶

Citation: Osborne JM, Bernabeu MO, Bruna M, Calderhead B, Cooper J, et al. (2014) Ten Simple Rules for Effective Computational Research. PLoS Comput Biol 10(3): e1003506. doi:10.1371/journal.pcbi.1003506

1. Look before you leap
2. Develop a prototype first
3. Make your code understandable to others (and yourself)
4. Don't underestimate the complexity of the task
5. Understand the mathematical, numerical, and computational methods underpinning your work
6. Use pictures: They really are worth a thousand words
7. Version control *everything*
8. Test *everything*
9. Share *everything*
10. Keep going!

Best Practices for Scientific Computing

Greg Wilson^{1*}, D. A. Aruliah², C. Titus Brown³, Neil P. Chue Hong⁴, Matt Davis⁵, Richard T. Guy⁶, Steven H. D. Haddock⁷, Kathryn D. Huff⁸, Ian M. Mitchell⁹, Mark D. Plumbley¹⁰, Ben Waugh¹¹, Ethan P. White¹², Paul Wilson¹³

Wilson G, Aruliah DA, Brown CT, Chue Hong NP, Davis M, et al. (2014) Best Practices for Scientific Computing. PLoS Biol 12(1): e1001745. doi:10.1371/journal.pbio.1001745

1. Write programs for people, not computers

- (a) keep program units small
- (b) use meaningful, distinctive, consistent names
- (c) make style and formatting consistent

2. Let the computer do the work

- (a) Make the computer repeat tasks
- (b) Save recent commands in file for re-use
- (c) Use a build tool to automate workflows

3. Make incremental changes

- (a) Work in small steps with frequent feedback and course correction (“agile”)
- (b) Use a version control system (VCS)
- (c) Version-control all manually created content

4. Don't repeat yourself (or others)

- (a) Every piece of data must have a single authoritative representation in the system
- (b) Modularize code (instead of copying and pasting)
- (c) Re-use code instead of rewriting it

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5. Plan for mistakes

- (a) add assertions to programs to check their operation
- (b) Use an off-the-shelf unit testing library
- (c) Turn bugs into test cases
- (d) Use a symbolic debugger

6. Optimize software only after it works correctly

- (a) Use a profiler to identify bottlenecks
- (b) Write code in the highest-level language possible

7. Document design and purpose, not mechanics

- (a) Document interfaces and reasons, not implementation
- (b) Refactor code in preference to explaining how it works
- (c) Embed the documentation in the software (and use documentation generators)

8. Collaborate

- (a) Use pre-merge code reviews
- (b) Use pair programming (bringing someone new up to speed, tricky problems)
- (c) Use an issue tracking tool

Basic software engineering

1. Version control
2. Automated testing
3. Documentation

Basic software engineering

1. Version control
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3. Documentation

<https://software-carpentry.org/>

SOFTWARE CARPENTRY

Getting Scientists to Write Better Code by Making Them More Productive

By Greg Wilson Computing In Science & Engineering, 8(6):66–69, 2006.

Version control with git

http://asu-compmethodphysics-phy494.github.io/ASU-PHY494/2018/01/30/04_Git_basics/

Why version control ?

"FINAL".doc



FINAL.doc!



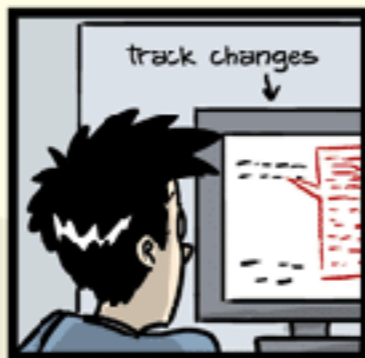
FINAL_rev.2.doc



FINAL_rev.6.COMMENTS.doc



FINAL_rev.8.comments5.
CORRECTIONS.doc



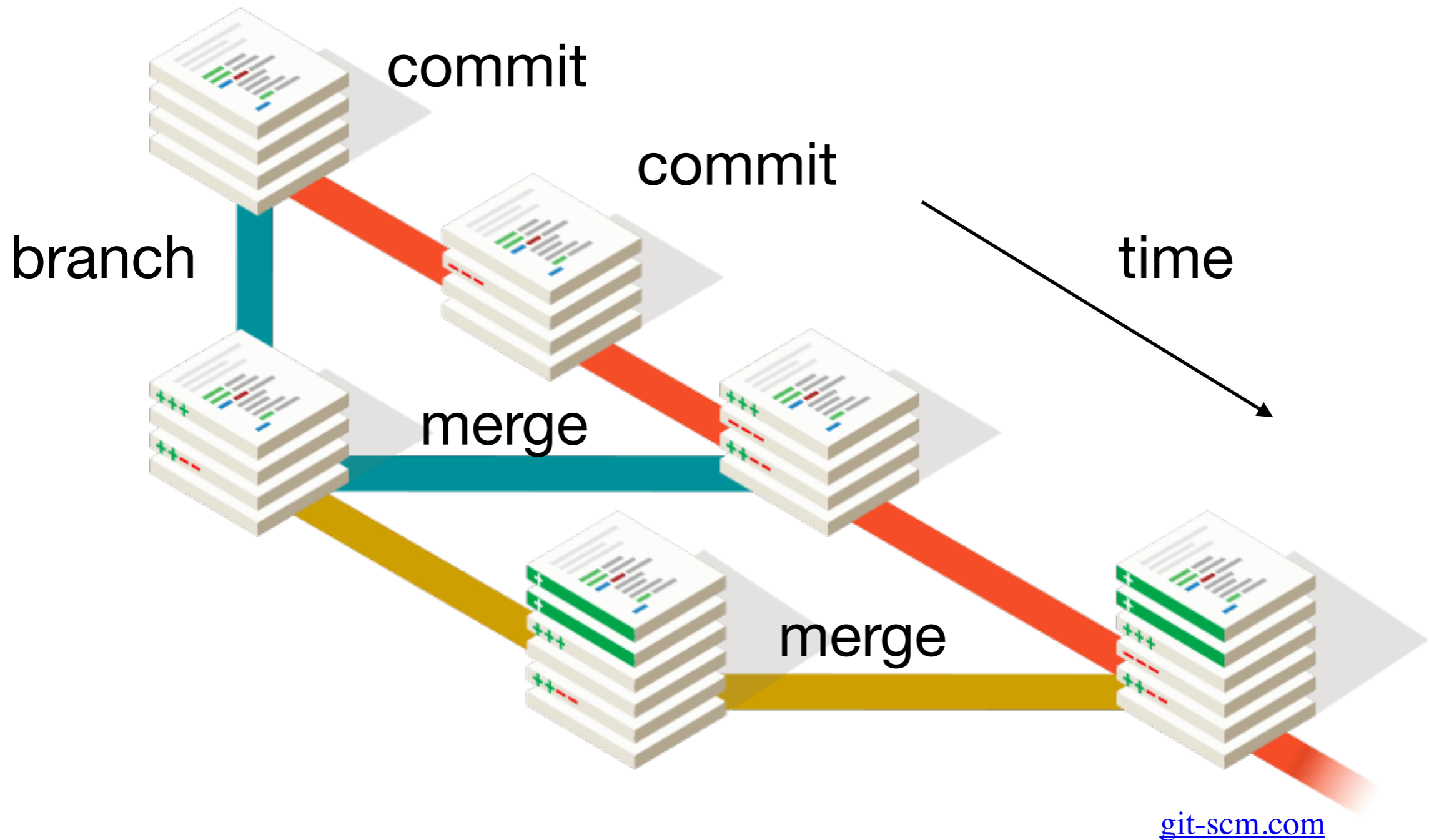
FINAL_rev.18.comments7.
corrections9.MORE.30.doc



FINAL_rev.22.comments49.
corrections.10.#@\$%WHYDID
ICOMETOGRADSCHOOL?????.doc

JORGE CHAM © 2012

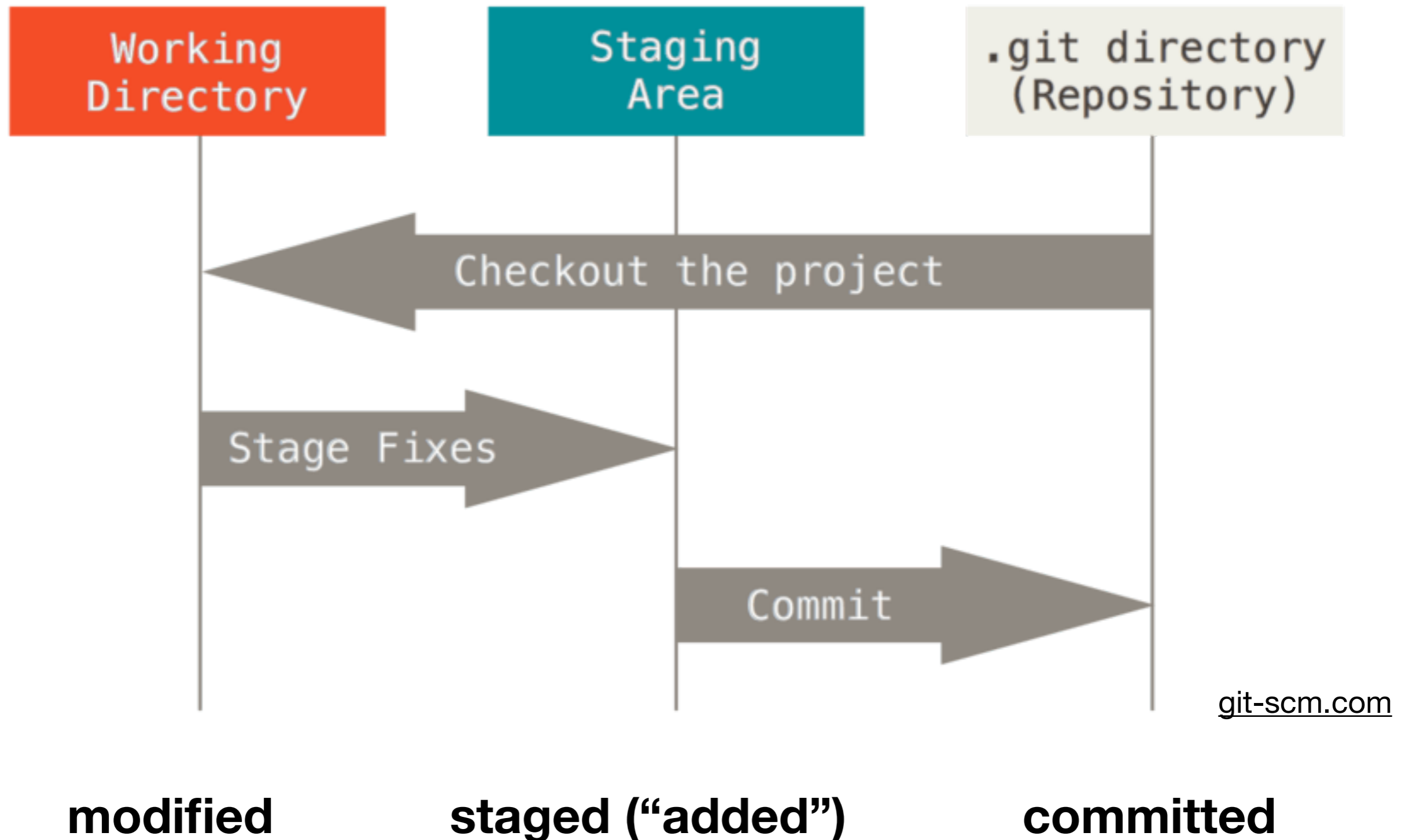
Version control with git



Create repository

```
cd mean_calculator  
git init
```

Stages of git



git workflow

1. **modify** files in working directory
2. **selectively stage changes** that you want to include in your next commit (adds only those files to the staging area)
3. **commit** your changes (takes files from the staging area and stores them permanently in your Git repository)

Check status

```
git status
```

Use this command gratuitously!

Add files

```
git add README.md *.py
```

Commit files

`git commit`

When your editor pops up, enter a **commit message**: Convention:

- first line (<60 char): one line summary
- second line: blank
- third and following lines: more details The first line is mandatory (you cannot have a commit without a message), the rest is optional. *The commit message should succinctly summarize the changes in the commit.*

Check status

```
git status
```

Use this command gratuitously!

git workflow

1. **modify** files in working directory
2. **selectively stage changes** that you want to include in your next commit (adds only those files to the staging area)

Adding files/changes: `git add`

Removing files: `git rm`

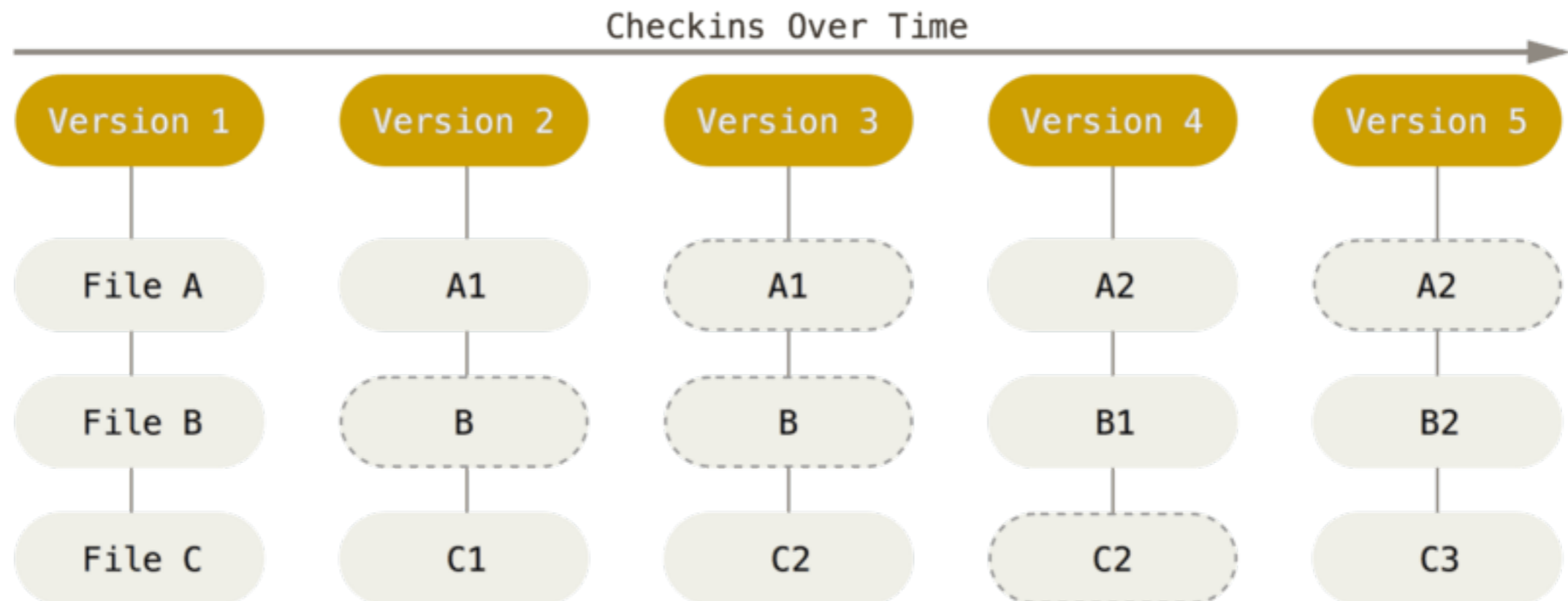
Renaming files: `git mv`

3. **commit** your changes (takes files from the staging area and stores them permanently in your Git repository)

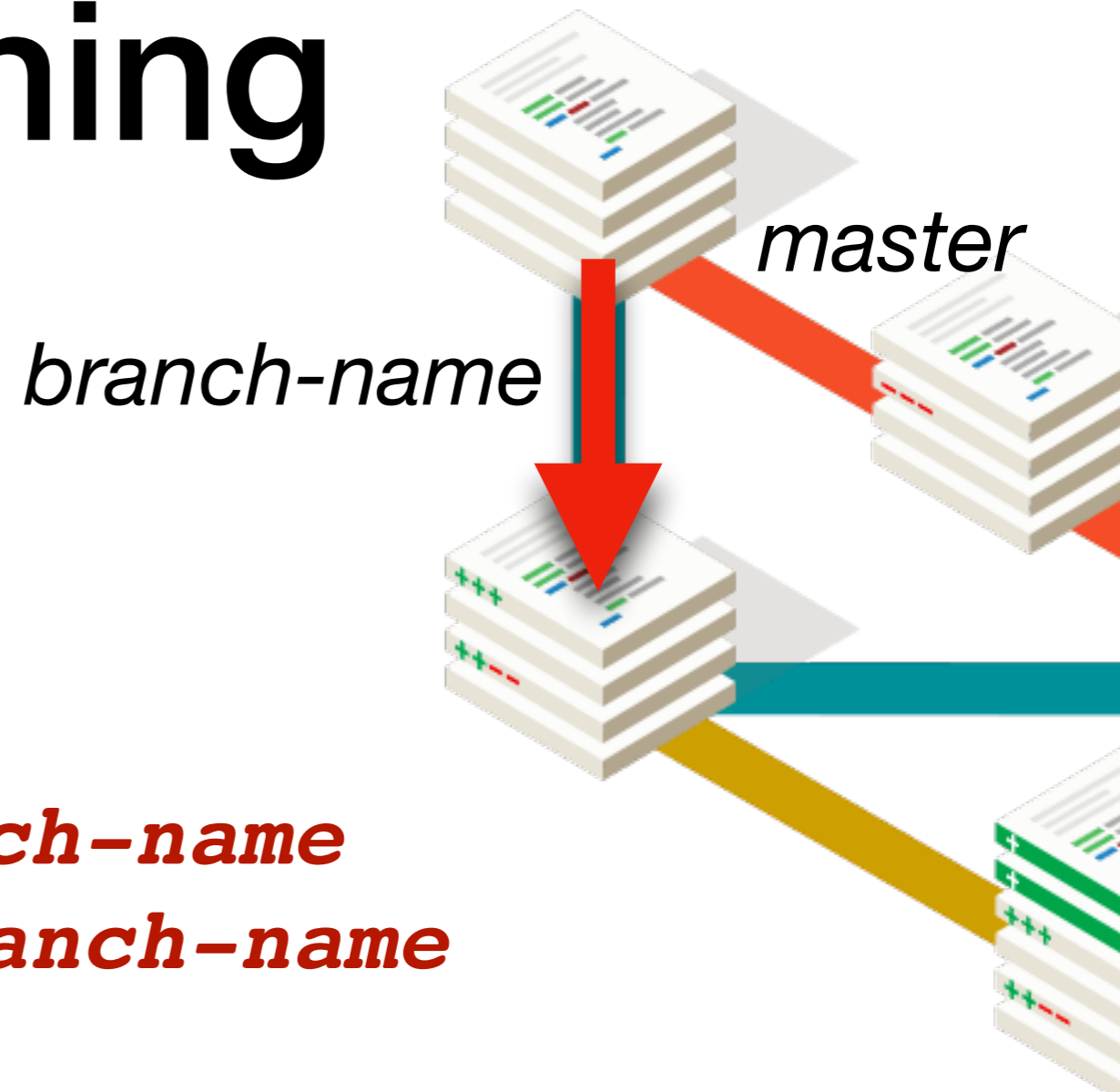
`git commit -m "message"`

History

`git log`



Branching



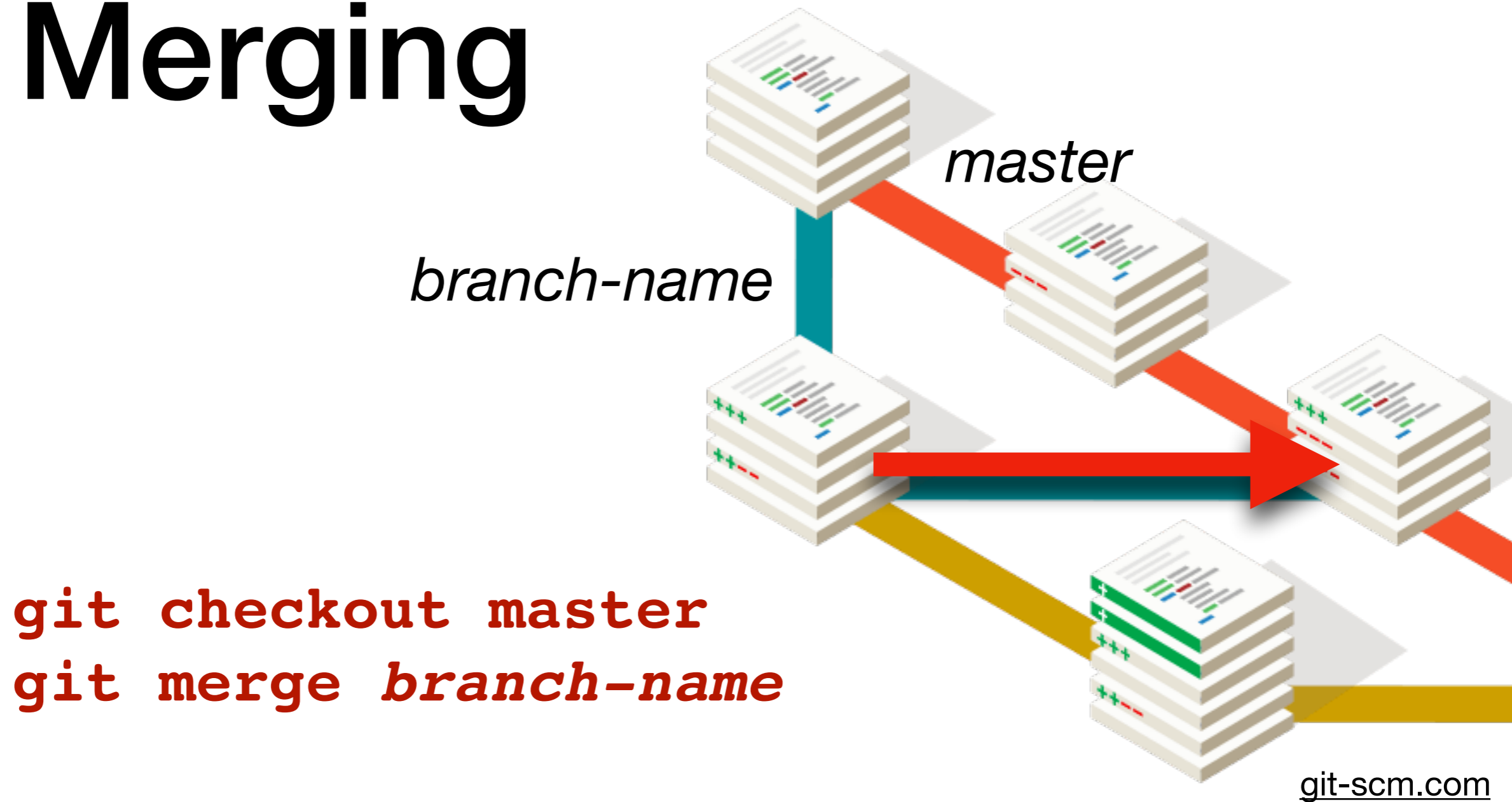
git branch *branch-name*
git checkout *branch-name*

git branch

git-scm.com

- Main branch: *master*
- Other branches: use meaningful names
- Short-cut: `git checkout -b branch-name`

Merging



- git tries automatic merging (and often succeeds)
- But... sometimes cannot reliably merge: **manually fix conflicts** (edit files: look for conflict markers <<<<, ==, >>>>)
- Read the output from **git merge** *very carefully!*

Remote repositories

```
git clone remote-repo-url  
cd repo
```

Create a local repository that is linked to the remote “origin”

```
git pull
```

Update local repository with remote content (“read”); merge if necessary.


```
git push
```

Update remote repository with local content (“write”); *must* pull first if some remote content is not present in local.

GitHub <https://github.com/>

- Cloud-based repositories
- Free for open source (and education)
- Create account and create new repo *mean_calculator*

Overview **Repositories 10** Stars 102 Followers 66 Following 11

 Type: All ▾ Language: All ▾  **New**

- Add LICENSE (later) and README
- Clone and add your files, then push

```
git clone https://github.com/YOUR_USER_NAME/mean_calculator
```

Testing

9/9

0800 Anttan started
1000 " stopped - anttan ✓
1300 (032) MP - MC $\left\{ \begin{array}{l} 1.2700 \quad 9.037847025 \\ 9.037846995 \text{ correct} \end{array} \right.$
~~1.482147000~~
~~2.130476415~~ (033) PRO 2 2.130476415
2.130676415 correct

Relays 6-2 in 033 failed special speed test
in relay " 10,000 test.

Relay
2145
Relay 3370

1100 Started Cosine Tape (Sine check)
1525 Started Multi-Adder Test.

1545



Relay #70 Panel F
(moth) in relay.

First actual case of bug being found.

~~1630~~ 1630 Anttan started.
1700 closed down.

Testing

9/9

0800 Antan started
1000 " stopped - antan ✓
1.2700 9.037 847 025
9.037 846 995 correct

Code without tests is legacy code.

1100 Started Cosine Tape (Sine check)
1525 Started Mult + Adder Test.

1545



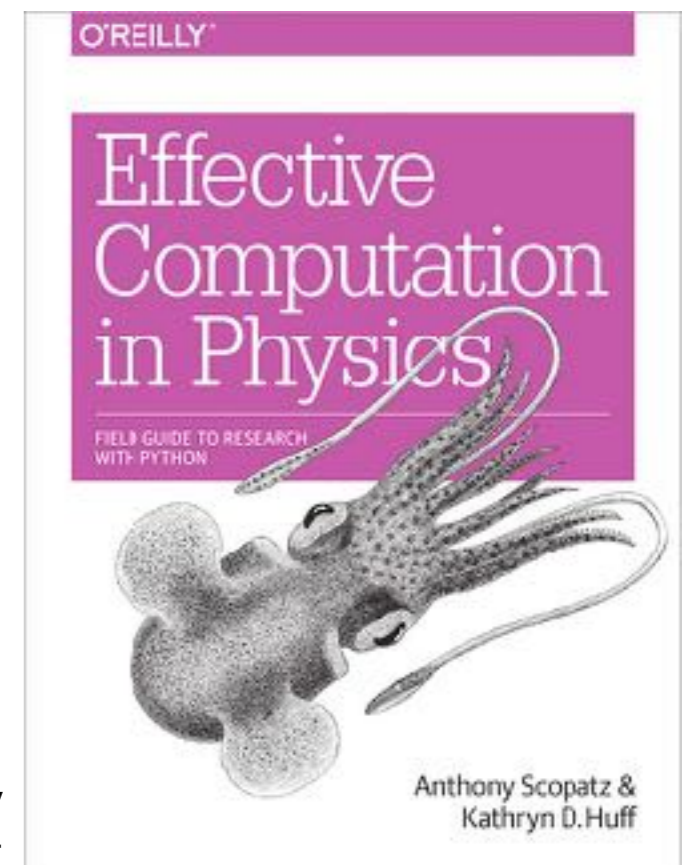
Relay #70 Panel F
(moth) in relay.

First actual case of bug being found.
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Tests

- Assert that your code produces known results.
- Tests are functions that
 - run your code
 - compare computed to known correct values
 - raise exception or return error if they disagree
- Write tests for
 - functions/methods/classes (unit tests)
 - modules/libraries (integration tests)
- Regression tests (compare to past values)

<https://katyhuff.github.io/python-testing/>



pytest

mean.py

```
def mean(num_list):  
    return sum(num_list)/len(num_list)
```

test_mean.py

```
import pytest  
  
from mean import mean  
  
def test_ints():  
    num_list = [1,2,3,4,5]  
    obs = mean(num_list)  
    exp = 3  
    assert obs == exp  
  
def test_zero():  
    num_list=[0,2,4,6]  
    obs = mean(num_list)  
    exp = 3  
    assert obs == exp
```

Run the tests

```
pytest
```

Continuous integration (CI)

- Does my software work on someone else's computer?
- With different versions of libraries/Python/ ...?

CI server

1. Checks out code from repository (triggered by push or pull request)
2. Spins up instances of operating systems (Linux, macOS, windows)
3. with required software versions (e.g., Python 2.7, 3.6)
4. Installs environment (libraries, ...)
5. Builds and installs software
6. Runs test scripts.
7. Checks for errors
8. Reports results (include coverage)

CI Providers

Providers with free plans (for open source)

Travis CI <https://travis-ci.com/> (Linux, macOS)

Appveyor <https://ci.appveyor.com/> (Windows)

Circle CI <https://circleci.com/> (Linux, macOS*)

...

Coverage reporting

codecov <https://codecov.io/>

coveralls <https://coveralls.io/>

Example: `pytest` + Travis CI

Based on Katy Huff's <https://kathyhuff.github.io/python-testing/08-ci/>

1. Create new repo on GitHub: *mean_calculator*
2. Clone locally
3. Add and commit example files from https://github.com/Becksteinlab/workshop_testing
4. Create account on <https://travis-ci.com> and allow *GitHub Apps Integration* to access all your repositories
5. Push changes (including the `.travis.yml` file): should trigger build on Travis-CI
6. check <https://travis-ci.com/> (when logged in, shows all your builds)

Documentation

- Code without documentation is close to useless (to others and to your future self).
- No-one likes writing documentation.

Getting docs done

- ➔ **Some/any documentation is better than none.**
 - ➔ Require docs in your projects.
 - ➔ Keep code and docs together (easier to write and maintain)
- ➔ **Use tools that make it *easy to generate docs*** (HTML, PDF, ...)
 - ➔ Document generators: *sphinx*, *doxygen*, ...
 - ➔ Human readable formats: restructured text (reST), markdown, ...
 - ➔ Automate doc creation: ReadTheDocs <https://readthedocs.org/>, GitHub pages <https://pages.github.com/> + CI



Star 219

Navigation

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Related Topics

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- Next: 1. Overview over MDAnalysis

Quick search

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Release: 0.18.1-dev

Date: Aug 05, 2018

MDAnalysis (www.mdanalysis.org) is an object oriented python toolkit to analyze molecular dynamics trajectories generated by [CHARMM](#), [Gromacs](#), [Amber](#), [NAMD](#), [LAMMPS](#), [DL_POLY](#) and other packages; It also reads other formats (e.g., [PDB](#) files and [XYZ format](#) trajectories; see [Table of supported coordinate formats](#) and [Table of Supported Topology Formats](#) for the full lists). It can write most of these formats, too, together with atom selections for use in [Gromacs](#), [CHARMM](#), [VMD](#) and [PyMol](#) (see [Selection exporters](#)).

It allows one to read molecular dynamics trajectories and access the atomic coordinates through [NumPy](#) arrays. This provides a flexible and relatively fast framework for complex analysis tasks. Fairly complete atom [Selection commands](#) are implemented. Trajectories can also be manipulated (for instance, fit to a reference structure) and written out in a range of formats.

Getting involved

Please report [bugs](#) or [enhancement requests](#) through the [Issue Tracker](#). Questions can also be asked on the [mdanalysis-discussion mailing list](#).

The MDAnalysis community subscribes to a [Code of Conduct](#) that all community members agree and adhere to – please read it.

Installing MDAnalysis

The easiest approach to [install the latest release](#) is to use a package that can be installed either with [pip](#) or [conda](#).

pip

Installation with [pip](#) and a *minimal set of dependencies*:

<https://gromacswrapper.readthedocs.io/en/develop/>

GromacsWrapper — a Python framework for Gromacs

Release: 0.6.1+70.g6b87aa8

Date: August 03, 2018

GromacsWrapper is a Python package that wraps system calls to [Gromacs](#) tools into thin classes. This allows for fairly seamless integration of the gromacs tools into [Python](#) scripts. This is generally superior to shell scripts because of Python's better error handling and superior data structures. It also allows for modularization and code re-use. In addition, commands, warnings and errors are logged to a file so that there exists a complete history of what has been done.

See [INSTALL](#) for download and installation instructions. [Documentation](#) is primarily provided through the Python doc strings (from which most of the online documentation is generated).

The source code itself is available in the [GromacsWrapper git repository](#).

Warning

Please be aware that this is **alpha** software that most definitely contains bugs. The API is not stable yet and can change between releases.

It is *your* responsibility to ensure that you are running simulations with sensible parameters.

The package and the documentation are still in flux and any [feedback](#), [bug reports](#), [suggestions](#) and contributions are very welcome. See the package [README: GromacsWrapper](#) for contact details.

See also



GromacsWrapper

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GromacsWrapper — a Python framework for Gromacs

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GromacsWrapper

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- [Admin](#)

Versions

latest	Public	Edit
master	Public	Edit
develop	Public	Edit

Repository

<https://github.com/Becksteinlab/GromacsWrapper>

Last Built

3 days, 9 hours ago passed

Maintainers



Badge

docs passing

Tags

Build a version

latest

[Build version](#)

Sharing code



Publish your computer code: it is good enough

Freely provided working code — whatever its quality — improves programming and enables others to engage with your research, says Nick Barnes.

14 October 2010 | Nature 467, 753 (2010) | doi:10.1038/467753a

I am a professional software engineer and I want to share a trade secret with scientists: **most professional computer software isn't very good.** The code inside your laptop, television, phone or car is often badly documented, inconsistent and poorly tested. [...]

That the code is a little raw is one of the main reasons scientists give for not sharing it with others. Yet, software in all trades is written to be good enough for the job intended. So **if your code is good enough to do the job, then it is good enough to release** — and releasing it will help your research and your field.

It is not common practice

People will pick holes and demand support and bug fixes.

The code is valuable intellectual property!

Too much work to polish code!

It should be...

Open-ness is the proper scholarly approach. Nobody is entitled to demand technical support for freely provided code: if the feedback is unhelpful, ignore it.

Rarely... almost all value is *your expertise*. Code not backed by experts = *abandonware*

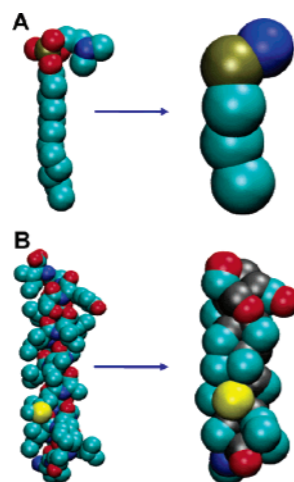
Does not have to be perfect — good enough is good!

Available = Citations

J|A|C|S
ARTICLES
Published on Web 02/04/2006

Insertion and Assembly of Membrane Proteins via Simulation

Peter J. Bond and Mark S. P. Sansom*



P. J. Bond and M. S. P. Sansom. Bilayer deformation by the Kv channel voltage sensor domain revealed by self-assembly simulations. *Proc Natl Acad Sci* 104(8): 2631–2636, 2007. **110 citations**

P. J. Bond and M. S. P. Sansom. Insertion and assembly of membrane proteins via simulation. *JACS* 128(8):2697–2704, Mar 2006. **331 citations**

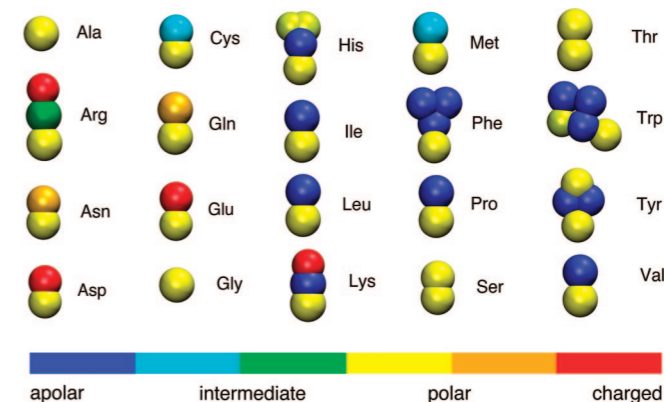
441 citations

J. Chem. Theory and Comput. 2008, 4, 819–834

JCTC Journal of Chemical Theory and Computation

The MARTINI Coarse-Grained Force Field: Extension to Proteins

Luca Monticelli,[†] Senthil K. Kandasamy,[‡] Xavier Periole,[§] Ronald G. Larson,[‡] D. Peter Tieleman,[†] and Siewert-Jan Marrink^{*,§}

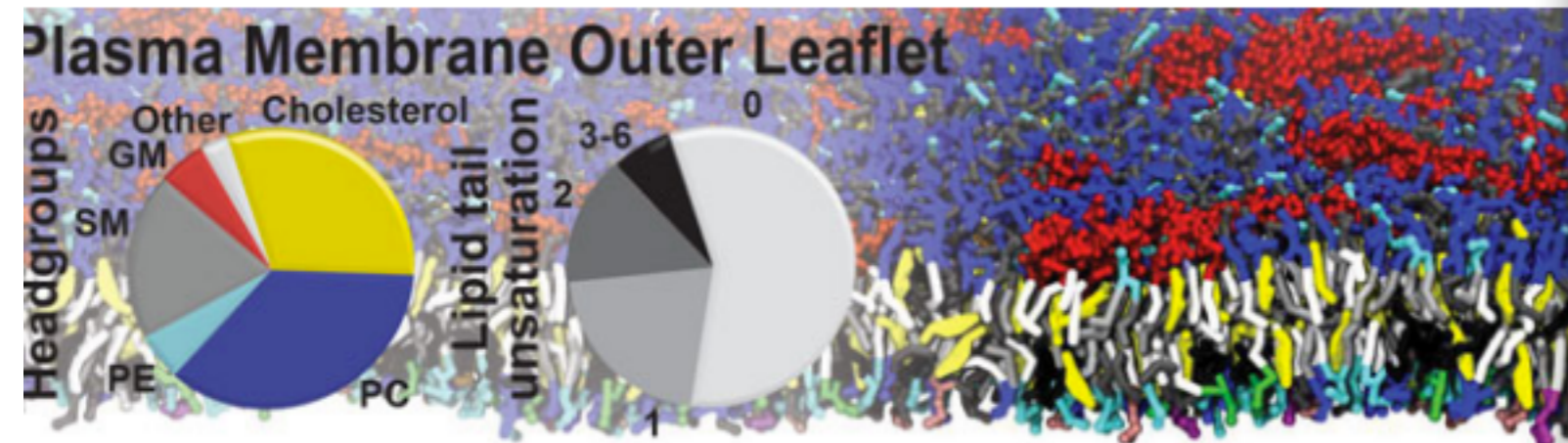


L. Monticelli, S. K. Kandasamy, X. Periole, R. G. Larson, D. P. Tieleman, and S.-J. Marrink. The MARTINI coarse-grained force field: Extension to proteins. *J Chem Theory Comput*, 4(5):819–834, 2008. **1476 citations**

1476 citations



Coarse Grain Forcefield for Biomolecules



J. Chem. Theory and Comput. **2008**, *4*, 819–834

JCTC Journal of Chemical Theory and Computation

The MARTINI Coarse-Grained Force Field: Extension to Proteins

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Download categories

Force field parameters

Example applications

Tools

Proteins and bilayers

Resolution transformation

Visualization

Proteins and bilayers

martinize

Last Updated: Thursday, 17 August 2017 11:59

Martimize is a python script to generate Martini protein topology and structure files based on an atomistic structure file. It replaces the old [seq2itp](#), [atom2cg](#) and [EINeDyn](#) scripts. The produced topology and structure files are in a format suitable for Gromacs.

<http://cgmartini.nl/index.php/tools2/proteins-and-bilayers>

Public repositories

Source code / VCS

- GitHub <https://github.com/>
- BitBucket <https://bitbucket.org/>
- SourceForge <https://sourceforge.net/>

Data / Source code (snapshot)

- Zenodo <https://zenodo.org/>
- figshare <https://figshare.com/>
- DataDryad <https://www.datadryad.org/>

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With material from and based on: *A short lecture on Open Licensing*, **Lorena A. Barba** (The George Washington University), <https://doi.org/10.6084/m9.figshare.4516892.v1>

See also: https://barbagroup.github.io/essential_skills_RRC/

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Education

A Quick Guide to Software Licensing for the Scientist-Programmer

Andrew Morin¹, Jennifer Urban², Piotr Sliz^{1*}

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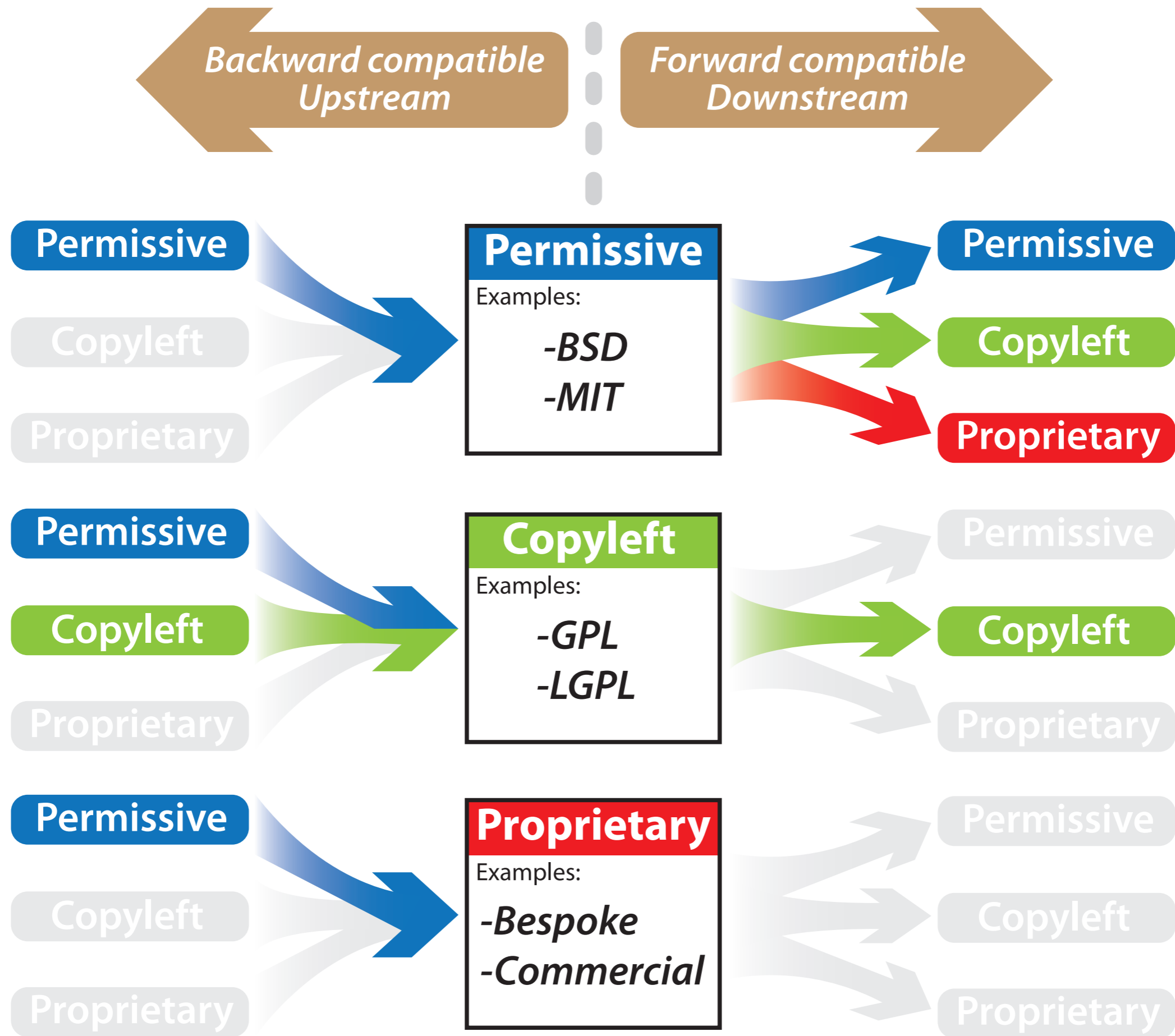
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Thank you!

- Links to material: <https://becksteinlab.physics.asu.edu/learning/117/summer-2018-mini-workshops>
- Videos: [YouTube: Becksteinlab Mini-Workshop 2018](#)