

# **Motivation and Overview of Best Practices in HPC Software Development**



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better scientific software Better Scientific Software tutorial @ SC21

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- The requested citation the overall tutorial is: David E. Bernholdt, Anshu Dubey, Patricia A. Grubel, Rinku K. Gupta, and Gregory R. Watson, Better Scientific Software tutorial, in the International Conference for High-Performance Computing, Networking, Storage, and Analysis (SC21), St. Louis, MO, USA and online, 2021. DOI: 10.6084/m9.figshare.16556628
- Individual modules may be cited as Speaker, Module Title, in Better Scientific Software tutorial...

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# Science through computing is, at best, as credible as the software that produces it!





The Success of Computational Science Creates the Challenges of Computational Science

- Positive feedback loop
  - More complex codes, simulations and analysis
  - More moving parts that need to interoperate
  - Variety of expertise needed the only tractable development model is through separation of concerns
  - It is more difficult to work on the same software in different roles without a software engineering process

More Hardware

Resources

- Onset of higher platform heterogeneity
  - Requirements are unfolding, not known a priori
  - The only safeguard is investing in flexible design and robust software engineering process

Supercomputers change fast Especially now!



Understanding

More Diverse

Solvers



**Higher Fidelity** 

Model

## **Challenges Developing Scientific Applications Today**

#### **Technical**

- All parts of the model and software system can be under research
- Requirements change throughout the lifecycle as knowledge grows
- Verification complicated by floating point representation
- Real world is messy, so is the software
- Increasing architectural diversity

## Sociological

- Competing priorities and incentives
  - Sponsors often care more about scientific publications than software per se
  - Balancing development and maintenance with research
- Limited resources
- Need for interdisciplinary interactions
  - Many different kinds of expertise to be successful

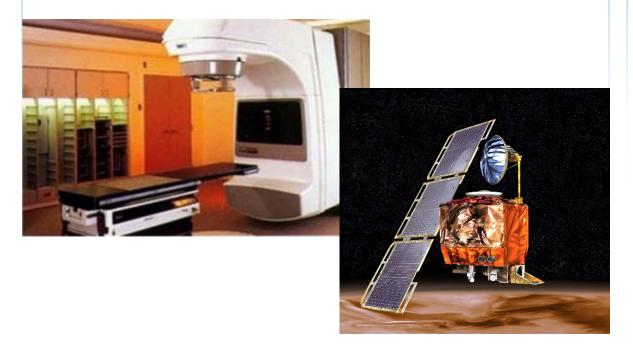




## High-Consequence Software-Related Scientific Failures

#### Therac-25 (1985-1987)

- Computer-controlled radiation therapy system
- Poor software design, development and testing practices allowed flaws that let to at least six cases of substantial radiation overdoses, three fatal



#### **Mars Climate Orbiter (1999)**

- Incorrect trajectory adjustment caused loss of the orbiter as it was supposed to enter Martian orbit
- Discrepancy in the units used in two different software components
- One component didn't follow specifications
- Inadequate testing at the interface
- Concerns raised earlier in the mission were ignored because they weren't properly documented

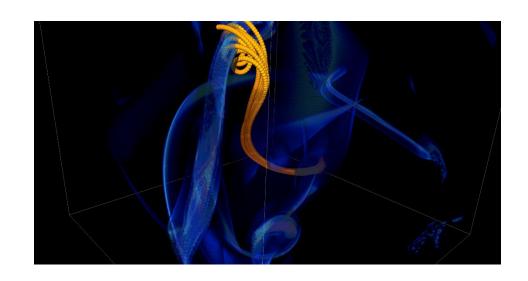
Just two of many examples





# More Subtle Impacts on Scientific Productivity

 In 2005, the FLASH astrophysics team was offered a unique opportunity to access one of the biggest machines in the world at that time (BG/L) for a dedicated run



- Short notice to prepare
  - < 1month to get ready for 1.5 week run</p>
- Quick and dirty development of particle capability in code
- Error in tracking particles resulted in duplicated tags from round-off
- Had to develop post-processing tools to correctly identify trajectories
  - 6 months to process results

FLASH had a software process in place. It was tested regularly. This was one instance when the full process could not be applied because of time constraints.

#### **Technical Debt**

The implied cost of additional rework caused by choosing an easy (limited) solution now instead of using a better approach that would take longer.

— Wikipedia

Like monetary debt, the more you accumulate, the harder it is to pay off

- Increases cost of maintenance
- Parts of software may become unusable over time
- Inadequately verified software produces questionable results
- Increases ramp-up time for new developers
- Overall, reduces software and science productivity





# Scientific Facilities Provide Valuable Resources

- Major supercomputers often cost O(\$100M)
- All cost millions more to operate, annually
- Significant allocations on large supercomputers can be worth millions
- Even if you don't pay the \$ you have to spend the time and effort to get the allocation
- <u>Sponsors' concern</u>: Are you being a good steward of the resources?
- Your concern: Are you getting the most science possible out of your work (aka scientific productivity)?











# Good scientific process requires good software practices



# Good software practices increase software sustainability







Software sustainability increases scientific productivity





### So, What Are Good Software Practices?

- There is no fixed, universally agreed set of best practices for scientific software
  - Specifics of what's appropriate will depend on the software, how it is used, and the team
- Let's look at a few recommendations from different perspectives...





# **Example 1: Best Practices for Scientific Computing (1/2)**

Wilson, et al., (2014) <a href="https://doi.org/10.1371/journal.pbio.1001745">https://doi.org/10.1371/journal.pbio.1001745</a>

- 1. Write programs for people, not computers.
  - a. A program should not require its readers to hold more than a handful of facts in memory at once.
  - b. Make names consistent, distinctive, and meaningful.
  - c. Make code style and formatting consistent.
- 2. Let the computer do the work.
  - a. Make the computer repeat tasks.
  - b. Save recent commands in a 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.
  - b. Use a version control system.
  - c. Put everything that has been created manually in version control.

- 4. Don't repeat yourself (or others).
  - a. Every piece of data must have a single authoritative representation in the system.
  - b. Modularize code rather than copying and pasting.
  - c. Re-use code instead of rewriting it.
- 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.





# **Example 1: Best Practices for Scientific Computing (2/2)**

Wilson, et al., (2014) <a href="https://doi.org/10.1371/journal.pbio.1001745">https://doi.org/10.1371/journal.pbio.1001745</a>

- 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 implementations.
  - b. Refactor code in preference to explaining how it works.
  - c. Embed the documentation for a piece of software in that software.

- 8. Collaborate.
  - a. Use pre-merge code reviews.
  - b. Use pair programming when bringing someone new up to speed and when tackling particularly tricky problems.
  - c. Use an issue tracking tool.





# Example 2: Good Enough Practices in Scientific Computing (1/2)

Wilson, et al., (2017) <a href="https://doi.org/10.1371/journal.pcbi.1005510">https://doi.org/10.1371/journal.pcbi.1005510</a>

#### 1.Data management

- a. Save the raw data.
- b. Ensure that raw data are backed up in more than one location.
- c. Create the data you wish to see in the world.
- d. Create analysis-friendly data.
- e. Record all the steps used to process data.
- f. Anticipate the need to use multiple tables, and use a unique identifier for every record.
- g. Submit data to a reputable DOI-issuing repository so that others can access and cite it.

#### 6. Manuscripts (out of order to save space)

- a. Write manuscripts using online tools with rich formatting, change tracking, and reference management.
- b. Write the manuscript in a plain text format that permits version control.

#### 2. Software

- a. Place a brief explanatory comment at the start of every program.
- b. Decompose programs into functions.
- c. Be ruthless about eliminating duplication.
- d. Always search for well-maintained software libraries that do what you need.
- e. Test libraries before relying on them.
- f. Give functions and variables meaningful names.
- g. Make dependencies and requirements explicit.
- h. Do not comment and uncomment sections of code to control a program's behavior.
- i. Provide a simple example or test data set.
- j. Submit code to a reputable DOI-issuing repository.





# Example 2: Good Enough Practices in Scientific Computing (2/2)

Wilson, et al., (2017) <a href="https://doi.org/10.1371/journal.pcbi.1005510">https://doi.org/10.1371/journal.pcbi.1005510</a>

#### Collaboration

- a. Create an overview of your project.
- b. Create a shared "to-do" list for the project.
- c. Decide on communication strategies.
- d. Make the license explicit.
- e. Make the project citable.

#### 4. Project organization

- a. Put each project in its own directory, which is named after the project.
- b. Put text documents associated with the project in the doc directory.
- c. Put raw data and metadata in a data directory and files generated during cleanup and analysis in a results directory.
- d. Put project source code in the src directory.
- e. Put external scripts or compiled programs in the bin directory.
- f. Name all files to reflect their content or function.

#### 5. Keeping track of changes

- a. Back up (almost) everything created by a human being as soon as it is created.
- b. Keep changes small.
- c. Share changes frequently.
- d. Create, maintain, and use a checklist for saving and sharing changes to the project.
- e. Store each project in a folder that is mirrored off the researcher's working machine.
- f. Add a file called CHANGELOG.txt to the project's docs subfolder.
- g. Copy the entire project whenever a significant change has been made.
- h. Use a version control system.





# Example 3: Linux Foundation Core Infrastructure Initiative (CII) Best Practices Badging Program

https://bestpractices.coreinfrastructure.org/en

- Not specifically intended for scientific software
- Three levels
  - **Passing** focuses on best practices that well-run FLOSS projects typically already follow. Getting the passing badge is an achievement; at any one time only about 10% of projects pursuing a badge achieve the passing level.
  - **Silver** is a more stringent set of criteria than passing but is <u>expected to be achievable by small and single-organization projects</u>.
  - Gold is even more stringent than silver and includes criteria that are not achievable by small or single-organization projects.
- Combination of MUST and SHOULD criteria





## **CII Best Practices Criteria Summary**

- Basics
  - Basic project website content (P, S)
  - FLOSS license (P)
  - Documentation (P, S)
  - Project oversight (S, G)
  - Accessibility and internationalization (S)
- Change control
  - Public version controlled source repo. (P, G)
  - Unique version numbering (P)
  - Release notes (P)
  - Previous versions (S)
- Reporting
  - Bug-reporting process (P, S)
  - Vulnerability reporting process (P, S)

(P, S, G) denotes additional criteria required at passing, silver, or gold certification levels

Each topic area listed will have one or more specific criteria

#### Quality

- Working build system (P, S, G)
- Automated test suite (P, S, G)
- New functionality testing (P, S)
- Warning flags (P, S)
- Coding standards (S, G)
- Installation system (S)
- Externally-maintained components (S)

#### Security

- Secure development knowledge (P, S)
- Use basic good crypto. practices (P, S, G)
- Secured delivery against MITM attacks (P, G)
- Publicly known vulnerabilities fixed (P)
- Secure release (S)

#### Analysis

- Static code analysis (P, S)
- Dynamic code analysis (P, S, G)





# Software Engineering Advice Often Needs Adaptation for Scientific Software

- The CII Best Practices are a good example of software engineering advice "in the wild"
- Experiences reported in the wild often don't consider the special nature of scientific software
- But that doesn't mean we should ignore all of the software engineering experience
  - Many useful concepts, approaches, and tools we can just <u>adopt</u>
- Some approaches may need to be adapted to work for scientific software
  - Find out how colleagues have addressed the challenges you're facing
    - Probably you will find multiple ways
  - In the end, some approaches may not work well
- Don't be afraid to experiment with adaptations
  - Consider using the PSIP process (coming up)





# How Much (Time, Effort) Should I Spend on Software Engineering?

Your project should include "just enough" software engineering so that you can meet your short-term and longer-term scientific goals effectively



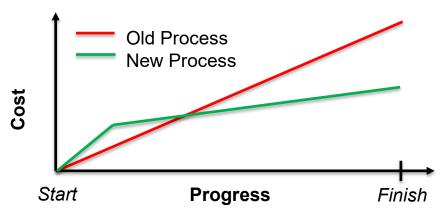


## **Continual, Incremental Software Process Improvement**

Target: your project should include "just enough" software engineering so that you can meet your short-term and longer-term scientific goals effectively

- 1. Identify your team's "pain points" in your software development processes
  - Help: RateYourProject assessment tool: <u>https://rateyourproject.org/</u>
- 2. Set a goal for something to improve
  - Target processes and behaviors, not just tasks
  - Pick something that you can address in a few months that will give you a noticeable benefit
- Agree on a plan to address it, identify markers of progress and what is "done"
  - Write them down
  - Help: Progress tracking card examples: https://bssw-psip.github.io/ptc-catalog/catalog
- 4. Work your plan, track your progress
- 5. When you are done, celebrate...

...then pick a new pain point to address



The new process costs something to implement, but it pays off over time

Productivity and Sustainability Improvement Planning <a href="https://bssw.io/psip">https://bssw.io/psip</a>



A goal of <u>BSSw.io</u> is to provide resources for improving your software processes. If you find useful resources that aren't on BSSw.io, consider contributing. Its easy and quick.





## **About Today's Tutorial**

- There are many useful topics that could help you improve your scientific software development process
- We're going to focus on a few where the software engineering advice in the wild typically doesn't address scientific software
  - Project management
  - Collaboration around software development
  - Designing software for flexibility and extensibility
  - Testing strategies for complex software systems
  - Systematic refactoring of large, complex software systems
  - Continuous integration testing
  - Reproducibility





# Agenda (Morning)

The agenda is also available on the tutorial web page. Visit <a href="https://bssw-tutorial.github.io">https://bssw-tutorial.github.io</a> and click on the link for today's tutorial

Time (CST)	Module	Title	Presenter
8:00 AM	0	Introduction and Setup	David E. Bernholdt (ORNL)
8:10 AM	1	Motivation and Overview of Best Practices in HPC Software Development	Patricia A. Grubel (LANL)
8:30 AM	2	Git Workflows	Patricia A. Grubel (LANL)
9:00 AM	3	Agile Methodologies	Rinku K. Gupta (ANL)
9:30 AM	4	Scientific Software Design	David E. Bernholdt (ORNL)
10:00 AM		Break	
10:30 AM	5	Agile Methodologies Redux	Rinku K. Gupta (ANL)
10:45 AM	6	Improving Reproducibility Through Better Software Practices	David E. Bernholdt (ORNL)
11:30 AM		Hands-on & Discussion	
12:00 PM		Lunch	



# Agenda (Afternoon)

The agenda is also available on the tutorial web page. Visit <a href="https://bssw-tutorial.github.io">https://bssw-tutorial.github.io</a> and click on the link for today's tutorial

Time (CST)	Module	Title	Presenter
12:00 PM		Lunch	
1:00 PM	7	Software Testing Introduction	Gregory R. Watson (ORNL)
1:30 PM	8	Continuous Integration	Gregory R. Watson (ORNL)
1:55 PM	9	Testing Complex Software	Anshu Dubey (ANL)
2:15 PM	10	Refactoring Scientific Software	Anshu Dubey (ANL)
3:00 PM		Break	
3:30 PM	11	Summary	David E. Bernholdt (ORNL)
3:45 PM		Hands-on & Discussion	
5:00 PM		Adjourn	



