



# Overview of Best Practices in HPC Software Development



Better Scientific Software Tutorial

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## Acknowledgements

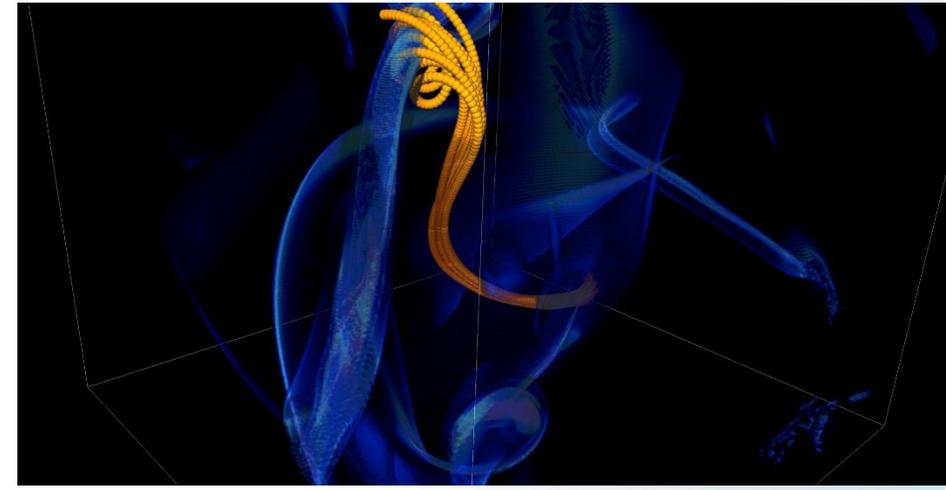
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**Good scientific process  
requires  
good software practices**

**Good software practices  
increase  
scientific productivity**

# You Can Mitigate Risk But It Is Never Zero

- Short notice availability of one of the biggest machines of it's time
  - **< 1month to get ready, run was 1.5 weeks**
- 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.

# Heroic Programming

Usually a pejorative term, is used to describe the expenditure of huge amounts of (coding) effort by talented people to overcome shortcomings in process, project management, scheduling, architecture or any other shortfalls in the execution of a software development project in order to complete it. Heroic Programming is often the only course of action left when poor planning, insufficient funds, and impractical schedules leave a project stranded and unlikely to complete successfully.

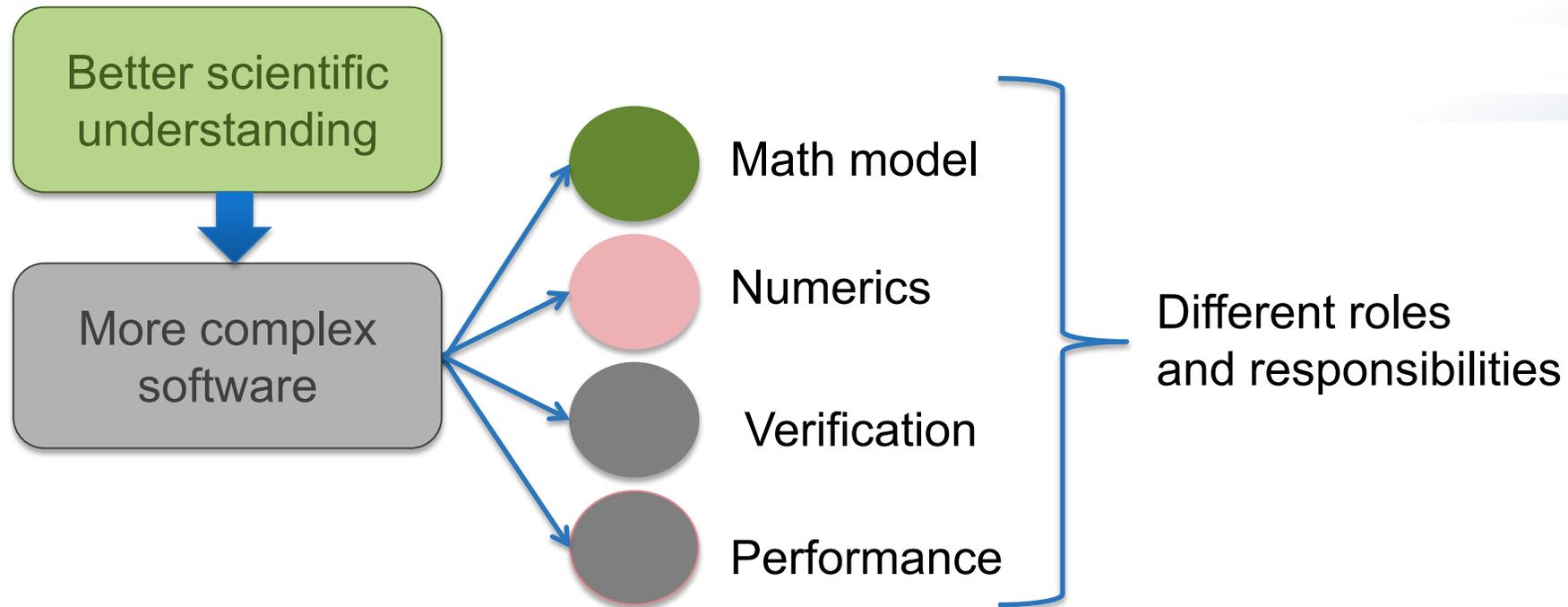
From <http://c2.com/cgi/wiki?HeroicProgramming>

**Science teams often resemble heroic programming**

Many do not see anything wrong with that approach

# What is wrong with heroic programming

Scientific results that could be obtained with heroic programming have run their course, because:



It is not possible for a single person to take on all these roles

# In Extreme-Scale science

- Codes aiming for higher fidelity modeling
  - 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**
- 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**

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Supercomputers change fast  
Especially Now

# Technical Debt

## Consequence of Choices

Quick and dirty collects interest which means more effort required to add features.

Accretion leads to unmanageable software

- Increases cost of maintenance
- Parts of software may become unusable over time
- Inadequately verified software produces questionable results
- Increases ramp-on time for new developers
- Reduces software and science productivity due to technical debt

# Challenges Developing a Scientific Application

## Technical

- All parts of the cycle 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

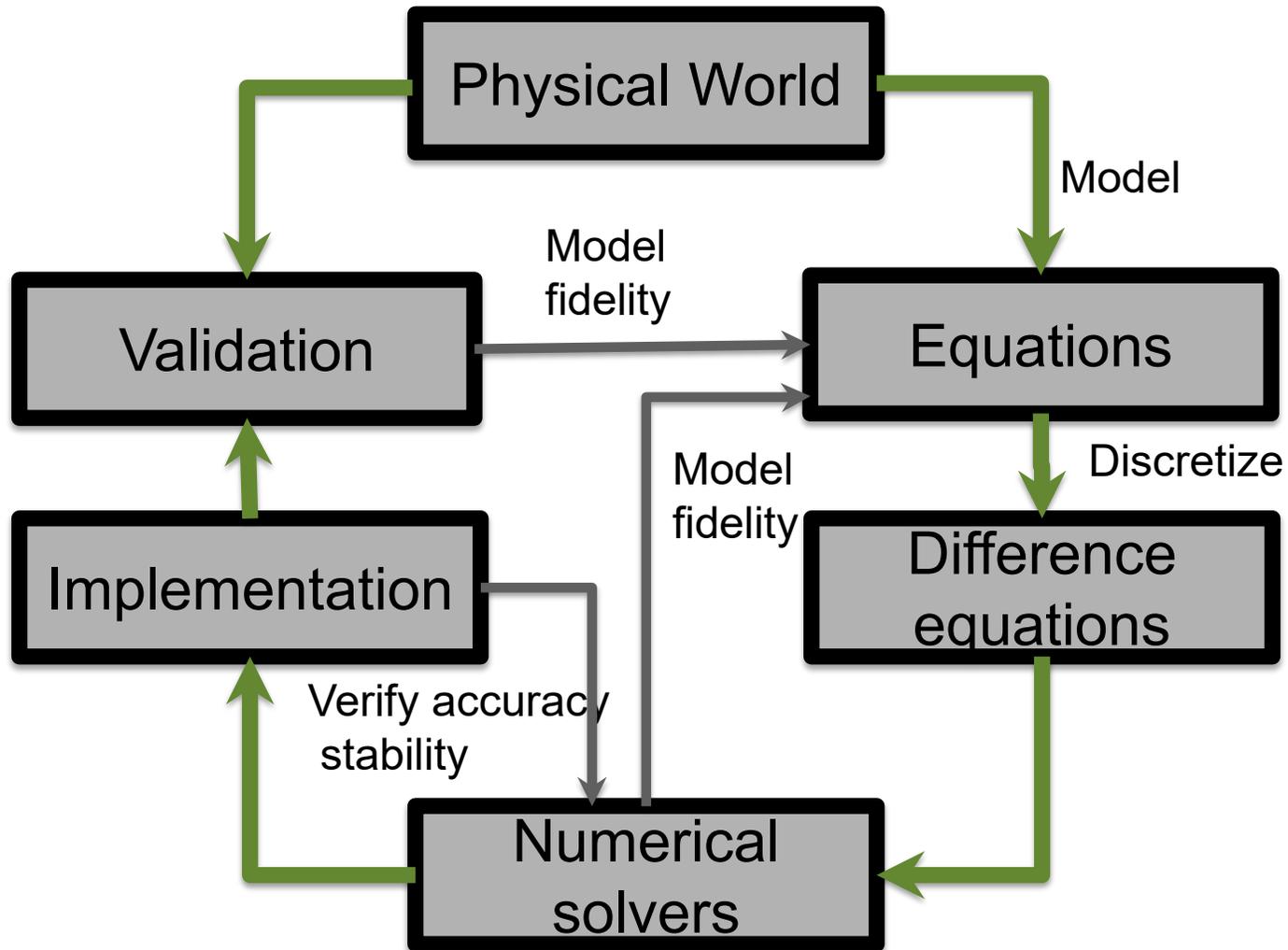
## Sociological

- Competing priorities and incentives
- Limited resources
- Perception of overhead without benefit
- Need for interdisciplinary interactions

# Customizations For Science Applications

- Testing does not follow specific methods as understood by the software engineering research community
  - The extent and granularity reflective of project priorities and team size
  - Larger teams have more formalization
- Lifecycle of science compare to lifecycle of development
- Development model
  - Mostly ad-hoc, some are close to agile model, but none follows it explicitly
  - Much more responsive to the needs of the lifecycle

# Lifecycle of Scientific Application



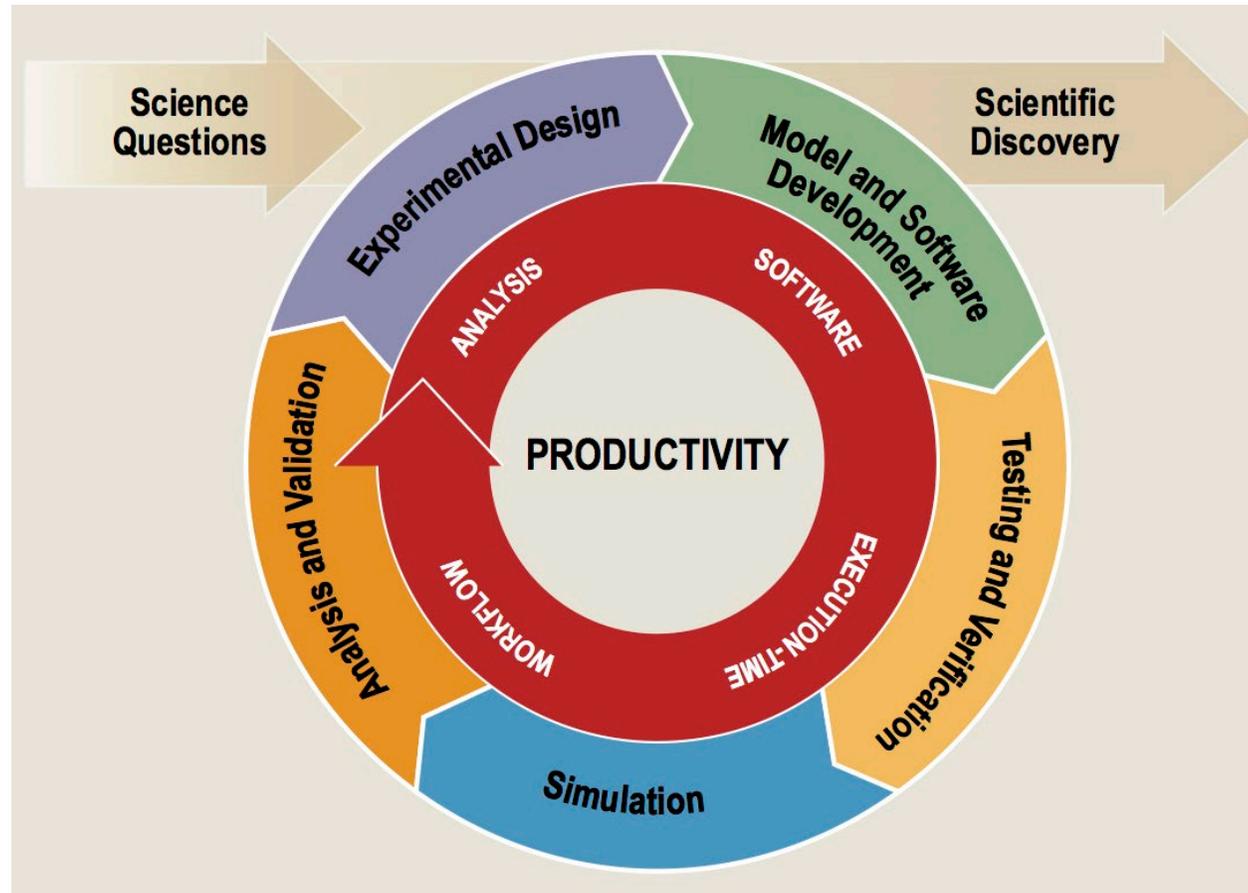
- Modeling

- Approximations
- Discretizations
- Numerics
  - Convergence
  - Stability

- Implementation

- Verification
  - Expected behavior
- Validation
  - Experiment/observation

# Software productivity cycle



<http://www.orau.gov/swproductivity2014/SoftwareProductivityWorkshopReport2014.pdf>

# Software Process Best Practices

## Baseline

- Invest in extensible code design
- Use version control and automated testing
- Institute a rigorous verification and validation regime
- Define coding and testing standards
- Clear and well defined policies for
  - Auditing and maintenance
  - Distribution and contribution
  - Documentation

## Desirable

- Provenance and reproducibility
- Lifecycle management
- Open development and frequent releases

# A Useful Resource

<https://ideas-productivity.org/resources/howtos/>

- **‘What Is’ docs:** 2-page characterizations of important topics for SW projects in computational science & engineering (CSE)
- **‘How To’ docs:** brief sketch of best practices
  - Emphasis on “bite-sized” topics enables CSE software teams to consider improvements at a small but impactful scale
- We welcome feedback from the community to help make these documents more useful

# Other resources

<http://www.software.ac.uk/>

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

<http://flash.uchicago.edu/cc2012/>

<http://journals.plos.org/plosbiology/article?id=10.1371/journal.pbio.1001745>

<http://ieeexplore.ieee.org/xpls/icp.jsp?arnumber=4375255>

<http://www.orau.gov/swproductivity2014/SoftwareProductivityWorkshopReport2014.pdf>

<http://ieeexplore.ieee.org/xpl/articleDetails.jsp?arnumber=6171147>

# Summary

- Good software practices are needed for scientific productivity
- Science at extreme-scales is complex and requires multiple expertise
- Software process does need to address reality
- Open codes, community contribution, are a powerful tool

It is extremely important to recognize that science through computing is only as good as the software that produces it