

# Software Testing and Verification

*Presented by*

**COLABS: Collaboration  
for Better Software for  
Science**

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Argonne National Laboratory

*In collaboration with*



*With prior support from*



Software Productivity and Sustainability track @ Argonne Training Program on Extreme-Scale Computing summer school


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- Individual modules may be cited as *Speaker, Module Title, in Tutorial Title, ...*

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# What is Testing

Whenever you write a code you are doing it

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- When you run it for the first time you are testing for correctness
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- When you break down your development into smaller chunks you test each chunk, then you combine the chunks, and you test again.

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Formalization of the process intimidates people because they think of writing tests as an overhead

# How to Think About Building Tests

You start by thinking about what is the correct behavior

Next you think about how you are going to be able to tell whether the code is exhibiting correct behavior

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Let us work through an example ...

- You want a large prime number for encryption
- As a part of the development, you first write a function that checks if a given number is prime

Correct behavior: input 13 returns true,  
input 15 returns false  
Incorrect behavior: input 15 returns true

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Here are all the ingredients for building a test !!

- You write a “main” that reads in a number, calls the functions and prints true or false
- You can automate it by including a series of known primes and non-primes and their corresponding true or false values
- This is your “unit test” for the function

# How to Think About Building Tests

Next you write a function to get to a large prime for encryption

Then you wish to confirm that it is a large enough prime

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Finally, you want to verify that it meets your encryption needs

You integrate your new function with your encryption software

The encryption software is likely to have a way to verify that the cipher can only be translated with the right key

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- Now you have a more complex test that involves several correctly working components
- This is your “integration test”

# Types of Tests

## Well known tests for enterprise software

- Unit tests – verify a single function, extremely quick to run
- Integration tests – verify functions working together
- System tests – verify functionality of the entire software
- Acceptance tests – verify that the client needs are met
- Regression tests – verify that there is no degradation in code capabilities

# Types of Tests

## **Additional types of tests needed for research software**

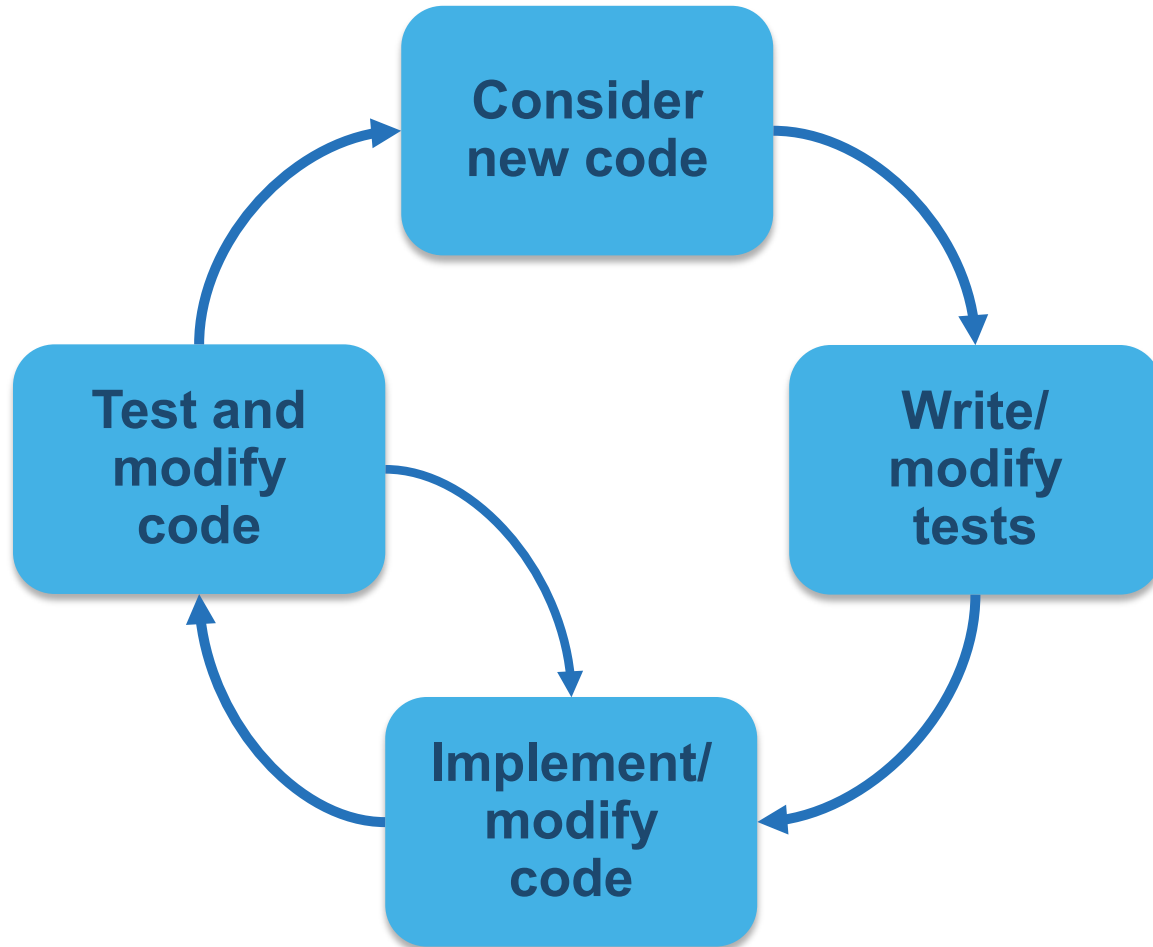
- Composite unit tests – are tests for specific functionalities and/or capabilities
- Granular tests – are integration tests at various granularities verifying correct behavior of interoperating functional units
- Restart tests – verify that a run can restart transparently from a checkpointed state
- Performance tests – apply to high-performance computing codes, verify that there is no performance loss



# Classes of Tests

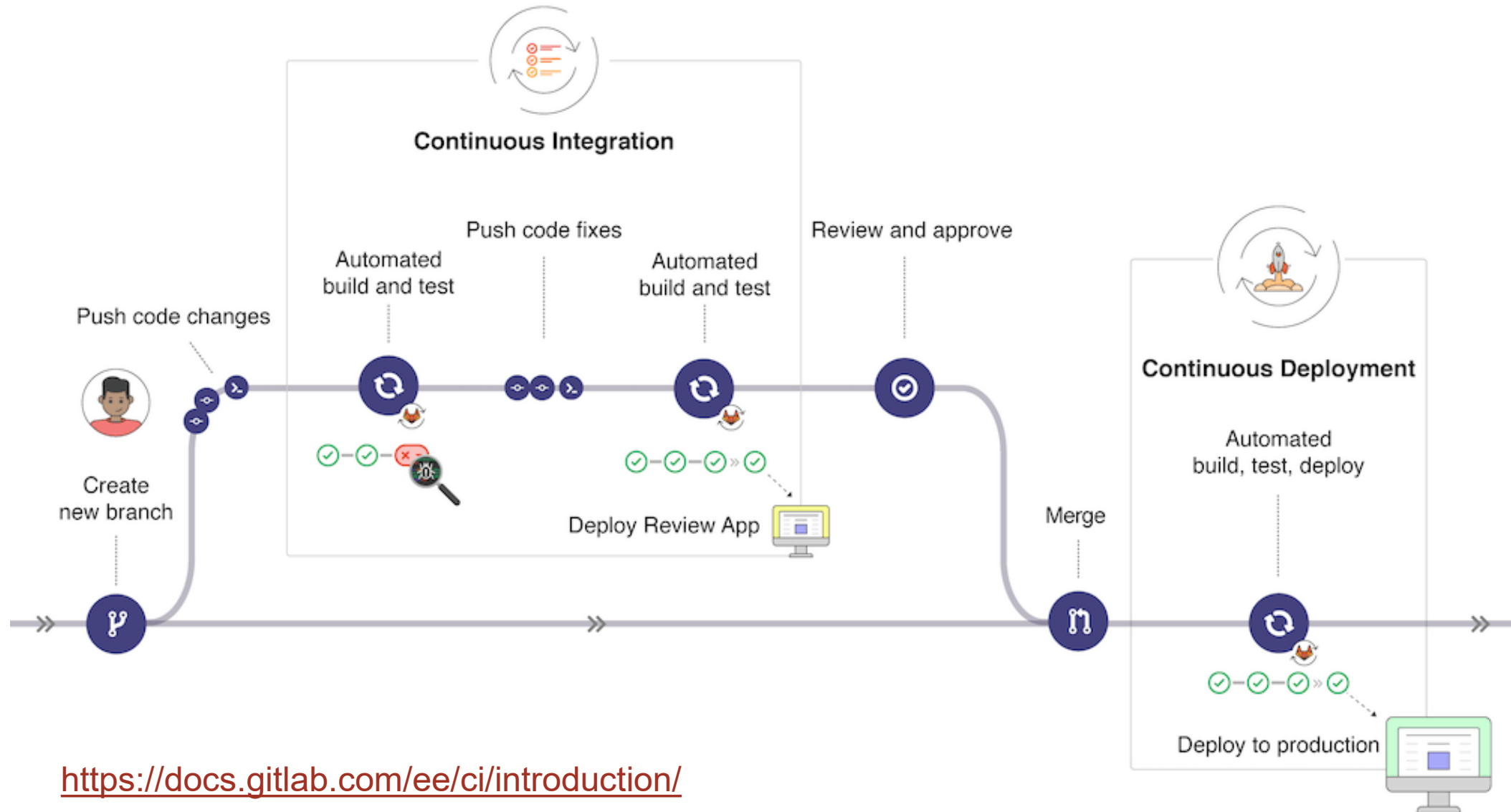
- White box testing – when you know the internals and can modify the code you are testing
  - Likely to be the code you and your collaborators are developing
  - You can insert assertions
  - You can insert code snippets that make testing easier
- Black box testing – when you do not know the internals of the code being tested, and cannot modify the code
  - Third party software or legacy code
  - The only means of verification available is reasoning about output to be obtained from supplied input

# Test Driven Development



- Documented specifications and requirements of the code
- Ensures that thought is given to what it means for the program to be correct, rather than just what the program should do
- More efficient development cycle
- Much less debugging
- Requires:
  - Care in writing tests
  - Frequent running of tests
  - Wide adoption by development team

# What is Continuous Integration (CI)



<https://docs.gitlab.com/ee/ci/introduction/>

# CI Components

- Testing
  - Focused, critical functionality (infrastructure), fast, independent, orthogonal, complete, ...
  - Existing test suites often require re-design/refactoring for CI
- Integration
  - Changes across key branches merged & tested to ensure the “whole” still works
    - Integration can take place at multiple levels
      - Individual project
      - Spack
      - E4S
    - Develop, develop, develop, merge, merge, merge, test, test, test...NO!
    - Develop, test, merge, develop, test, merge, develop, test, merge...YES!
- Continuous
  - Changes tested every commit and/or pull-request (like auto-correct)
- CI generally implies a lot of automation

# Test Driven Development vs. Automated Testing vs. CI

- ***Test Driven Development***: A development methodology where functional test are written before the code
  - Works well with CI as tests are written and committed and are automatically run (failing)
  - Code that implements the functionality being tested retriggers the tests automatically
- ***Automated Testing***: Software that automatically performs tests on a regular basis and reliably detects and reports anomalous behaviors/outcomes.
  - Examples: Auto-test, CTest/CDash, nightly testing, etc.
  - May live “next to” your development workflow
  - Potential issues: change attribution, timeliness of results, multiple branches of development
- ***Continuous Integration (CI)***: automated testing performed at high frequency and fine granularity
  - Aimed at preventing code changes from breaking key branches of development (e.g. main)
  - Lives “within” your development workflow
  - Potential issues: extreme automation, test granularity, coverage, 3<sup>rd</sup>-party services/resources

# Examples...

Automated Nightly Testing Dashboard  
Lives “next to” your development work



CI Testing  
Lives embedded in your development work



## Results of Visit Regression Test ( pascal,trunk,serial )



Test suite run started at 2020:07:09:22:49:46.  
(Click on table header to sort)


Index	Category	Test File	Status	Runtime (sec)
243	rendering	<a href="#">ospray.py</a>	Unacceptable	5.0
273	simulation	<a href="#">batch.py</a>	Unacceptable	38.0
24	databases	<a href="#">chgcarr.py</a>	Succeeded With Skips	11.0
32	databases	<a href="#">exodus.py</a>	Succeeded With Skips	14.0
66	databases	<a href="#">silo.py</a>	Succeeded With Skips	50.0
67	databases	<a href="#">silo_altdriver.py</a>	Succeeded With Skips	87.0
75	databases	<a href="#">xdmf.py</a>	Succeeded With Skips	14.0
109	hybrid	<a href="#">merge_tree.py</a>	Succeeded With Skips	11.0
136	meshtype	<a href="#">emptydomains.py</a>	Succeeded With Skips	7.0
256	rendering	<a href="#">view.py</a>	Succeeded With Skips	17.0
275	simulation	<a href="#">curve.py</a>	Succeeded With Skips	8.0
281	simulation	<a href="#">life.py</a>	Succeeded With Skips	8.0
296	simulation	<a href="#">zerocopy.py</a>	Succeeded With Skips	32.0
0	databases	<a href="#">ANALYZE.py</a>	Succeeded	10.0
1	databases	<a href="#">ANSYS.py</a>	Succeeded	9.0
2	databases	<a href="#">CGNS.py</a>	Succeeded	11.0
3	databases	<a href="#">Cale.py</a>	Succeeded	6.0
4	databases	<a href="#">Chombo.py</a>	Succeeded	7.0
5	databases	<a href="#">EnSight.py</a>	Succeeded	9.0
6	databases	<a href="#">FITS.py</a>	Succeeded	8.0
7	databases	<a href="#">Fluent.py</a>	Succeeded	7.0
8	databases	<a href="#">GDAL.py</a>	Succeeded	20.0
9	databases	<a href="#">NASTRAN.py</a>	Succeeded	15.0

Add more commits by pushing to the `edit-patch-1` branch on `edit/chromium-dashboard`.


  **All checks have passed** [Hide all checks](#)  
2 successful checks

  **Lighthouse** — Passed. New Lighthouse score would be 100/100. [Details](#)

  **continuous-integration/travis-ci/pr** — The Travis CI build passed [Details](#)

 **This branch has no conflicts with the base branch**  
Merging can be performed automatically.

[Merge pull request](#) You can also [open this in GitHub Desktop](#) or view [command line instructions](#).

 Write Preview AA B i “ <> 🔗 ☰ ☰ ☰ ↶ @ 📄

# What can make CI difficult

## Common situations

- Just getting started
  - Many technologies/choices; often in the "cloud"
  - Solution: start small, simple, build up
- Developing suitable tests
  - Many project's existing tests not suitable for CI
  - CI testing is a balance of thoroughness and responsiveness
  - Solution: Simplify/refactor and/or sub-setting test suite
- Ensuring sufficient coverage
  - Some changes to code never get tested – CI can provide a false sense of security
  - Solution: tools to measure it, enforce always increasing

## Advanced situations

- Defining failure for *many* configurations / inconsistent failures
  - Bit-for-bit (exact) match vs. fuzzy match
  - Solution: absolute/relative tolerances → AI/ML
- Numerous 3<sup>rd</sup> party libraries (TPLs)
  - Compiling takes too long
  - Solution: cache pre-built TPLs, containers
- Performance testing
  - Avoid time-, space-, scaling-performance degradation
  - Solution: Performance instrumentation and *scheduled* testing

# CI Resources (Where do jobs run?)

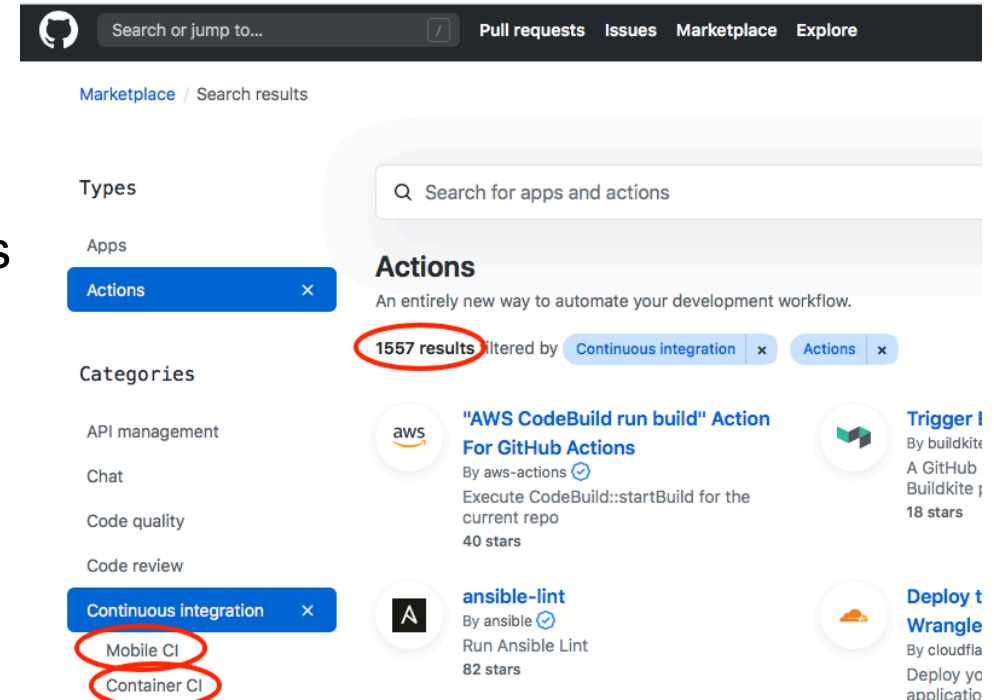
- Free Resources

- GitHub, BitBucket, GitLab, etc. provide shared runners
- AWS, Azure Pipelines have free tiers that can be used
- All launch a VM (Linux variants, Windows and OSX)
  - Constrained in time/size, hardware (e.g. GPU type/count)
  - Not a complete solution for many HPC/scientific codes, but a useful starting point.

- Site-local Resources

- Group, department, institution, computing facility
- Examples: CADES @ ORNL, Bamboo @ LLNL, Jenkins @ ANL, Travis+CDash @ NERSC
- ECP Program: GitLab-CI @ ANL, LANL, LLNL, NERSC, ORNL, SNL

- Create your own by setting up resources/services





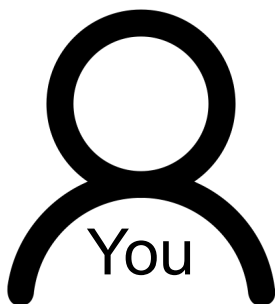
# Examples...



Your code repository



Your CI Resources



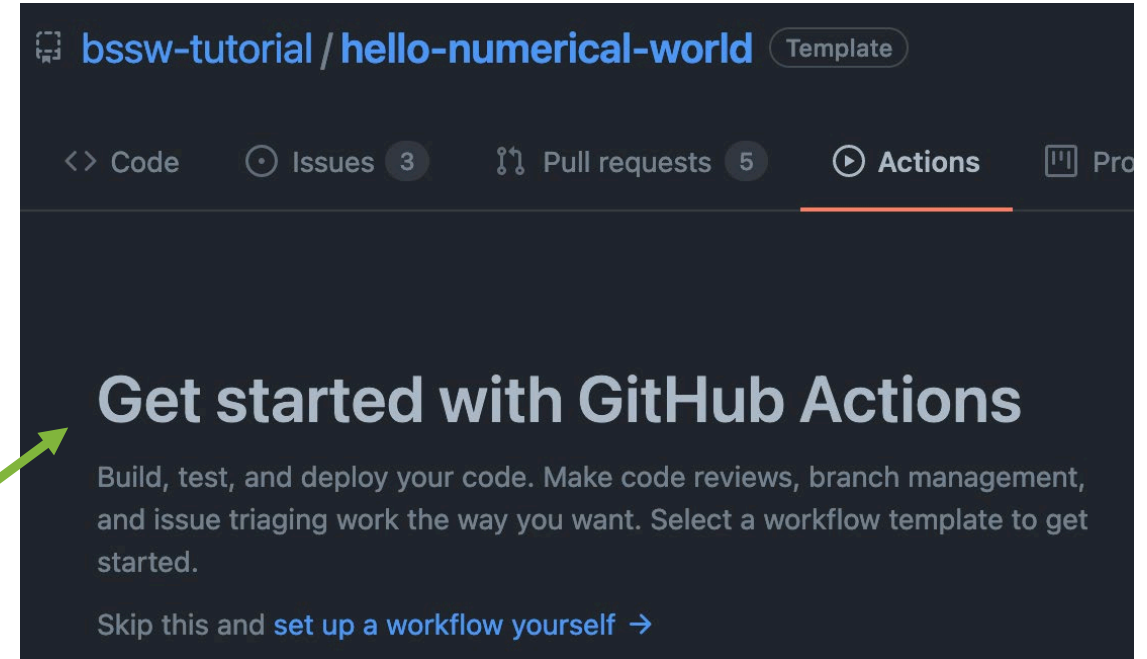
# Getting started with CI

- What *configuration* is most important?
  - Examples: gcc, icc, xlc? MPI-2 or MPI-3? Python 2, 3 or 2 & 3?
- What *functionality* is most important?
  - Examples: vanilla numerical kernels? OpenMP kernels? GPU kernels? All of these?
- Good candidates...
  - A “hello world” example for your project
  - At a minimum, even just building the code can be a place to start!
  - Once you’ve got the basics working, its easy to build up from there

# Getting started with CI:

## Setting up CI

Service	Interface	
GitHub Actions	Repo YAML file	<code>.github/workflows/&lt;test_name&gt;.yml</code>
GitLab	Web page configurator + repo YAML file [& repo scripts]	<code>/.gitlab-ci.yml</code> in root of repo
Bamboo	Web page configurator + repo scripts	
Travis	repo YAML file [& repo scripts]	<code>/.travis.yml</code> in root of repo



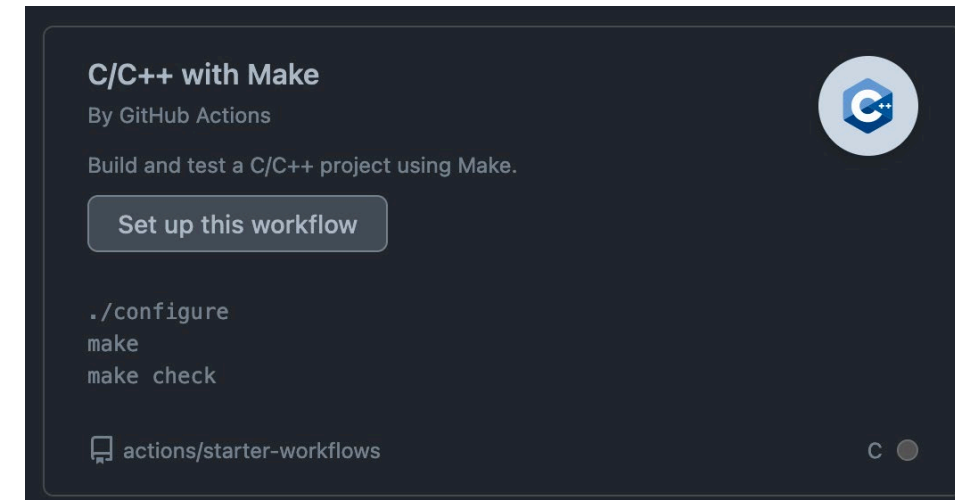
bssw-tutorial / hello-numerical-world Template

<> Code Issues 3 Pull requests 5 Actions

## Get started with GitHub Actions

Build, test, and deploy your code. Make code reviews, branch management, and issue triaging work the way you want. Select a workflow template to get started.

Skip this and [set up a workflow yourself](#) →



**C/C++ with Make**  
By GitHub Actions

Build and test a C/C++ project using Make.

[Set up this workflow](#)

```
./configure  
make  
make check
```

actions/starter-workflows

# Getting started with GitHub Actions:

19 lines (15 sloc) | 359 Bytes

```
1  name: Check Results
2
3  on:
4    push:
5      branches: [ main ]
6    pull_request:
7      branches: [ main ]
8
9  jobs:
10   build:
11
12     runs-on: ubuntu-latest
13
14     steps:
15     - uses: actions/checkout@v2
16     - name: make coverage
17       run: make CXXFLAGS=--coverage LDFLAGS="--coverage -lm" check_all
18     - name: upload coverage
19       run: bash <(curl -s https://codecov.io/bash)
```

# github.com

The screenshot shows the GitHub Actions interface for a workflow named 'build'. The workflow has succeeded 4 minutes ago. The 'make coverage' step is expanded, showing the following terminal output:

```
1 ▶ Run make CXXFLAGS=--coverage LDFLAGS="--coverage -lm"
   check_all
4 g++ -c --coverage heat.C -o heat.o
5 g++ -c --coverage utils.C -o utils.o
6 g++ -c --coverage args.C -o args.o
7 g++ -c --coverage exact.C -o exact.o
```

# codecov.io

The screenshot shows the Codecov.io coverage report for a pull request titled 'fix error threshold'. The overall coverage is 51.60%. The report includes a table of file coverage:

Files	Coverage
<a href="#">Double.H</a>	65.63%
<a href="#">args.C</a>	82.05%
<a href="#">crankn.C</a>	0.00%
<a href="#">exact.C</a>	0.00%
<a href="#">ftcs.C</a>	100.00%
<a href="#">heat.C</a>	73.81%
<a href="#">upwind15.C</a>	0.00%
<a href="#">utils.C</a>	49.35%
<b>Project Totals</b> (8 files)	<b>51.60%</b>

# GitHub Actions – results of workflow test runs

## Workflows

All workflows

- (TEST) Pyomo Windows Tests ...
- (WIP) Pyomo Windows Test (P...
- (WIP) Pyomo Windows Test (P...
- (WIP) Pyomo Windows Tests (...)
- (WIP) Windows Pip Cmd Pyom...
- GitHub Branch CI
- GitHub CI
- Pyomo Release Distribution Cr...
- Python package
- Ubuntu Pyomo Single Python ...
- Ubuntu Pyomo Workflow (Slim,...

## GitHub CI

Showing runs from all workflows named GitHub CI

event:push workflow:"GitHub CI"

461 workflow run results

Event ▾

Status ▾

Branch ▾

Actor ▾

✓	<b>Merge pull request #1902 from jsiiro/fix-unittest-rc</b> GitHub CI #121: Commit 901b487 pushed by blnicho	main	📅 20 hours ago 🕒 1h 3m 55s	...
✓	<b>Merge pull request #1901 from mrmundt/remove-six</b> GitHub CI #117: Commit a101b6d pushed by mrmundt	main	📅 2 days ago 🕒 1h 3m 12s	...
✓	<b>Merge pull request #1896 from jsiiro/abstract-disa...</b> GitHub CI #112: Commit 1f9dd19 pushed by jsiiro	main	📅 2 days ago 🕒 1h 5m 39s	...
✗	<b>Merge pull request #1898 from mrmundt/py-unittest</b> GitHub CI #109: Commit 1beb848 pushed by michaelbynum	main	📅 3 days ago 🕒 1h 3m 33s	...
✓	<b>Merge pull request #1893 from jsiiro/config-enum</b> GitHub CI #105: Commit 9aa1186 pushed by jsiiro	main	📅 3 days ago 🕒 1h 11m 3s	...

# GitHub Pull Request Status Indicators

Filters ▾	is:pr is:open	Labels 26	Milestones 3	New pull request			
🔗 17 Open ✓ 1,018 Closed	Author ▾	Label ▾	Projects ▾	Milestones ▾	Reviews ▾	Assignee ▾	Sort ▾
🔗 <b>Allow string args to external fuctions</b> ✖					ⓘ 1		💬 1
#1904 opened 19 hours ago by eslickj • Review required							
🔗 <b>Port Cloning</b> ✓					ⓘ 1		💬 3
#1899 opened 3 days ago by michaelbynum • Approved							
🔗 <b>Remove pyomo.solvers.tests.core</b> ✓							💬 2
#1897 opened 3 days ago by mrmundt • Approved							
🔗 <b>[WIP] Model partition package</b> ●							💬 1
#1895 opened 4 days ago by rahuljoglekar47 • Review required							

# What is CI Good For

- The purpose of CI is to identify problems early
  - Prevent code that would “break the build” or adversely impact other developers being introduced
  - Need to provide sufficient confidence, but run quickly – balance varies by project
- CI should complement (not replace) more extensive automated testing
  - Use scheduled testing for more and more detailed tests, more configurations and platforms, performance testing, etc.
- CI for TDD is a natural fit
  - Writing tests before the code works well with CI
- Many options for where to execute CI tests
  - Free services are a good (easy) place to start
  - But may not be sufficient in the long run (especially large HPC/scientific codes)
- Start simple to get automation working, then build out what you need
  - Focus initially on key software configurations and aspects of the code to be tested
  - Make sure your testing expands to cover new code, use TDD



# Building a Test-suite

## Elements of test development

- For some tests assertions will suffice
- For others you will need to compare the output against baselines
  - Building a comparison utility is extremely useful
- Also useful to develop diagnostics – indirect ways of verifying behavior
  - Conservation of physical quantities
  - No non-physical values

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## Building baselines for comparison

- From a known analytical solution
- Manufacture a solution
- Visualize and inspect output and anoint as baseline
- Run a test case up to point A and drop a checkpoint. Run another test case up to a later point B.
  - Use point A to restart and B as the anointed baseline

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Apply scaffolding  
for selection of  
tests ...  
explained next

# Example – Shock Hydrodynamics with Adaptive Mesh Refinement

## Components needed

- Mesh
- Hydrodynamics solver
- Equation of state
- Parallelization

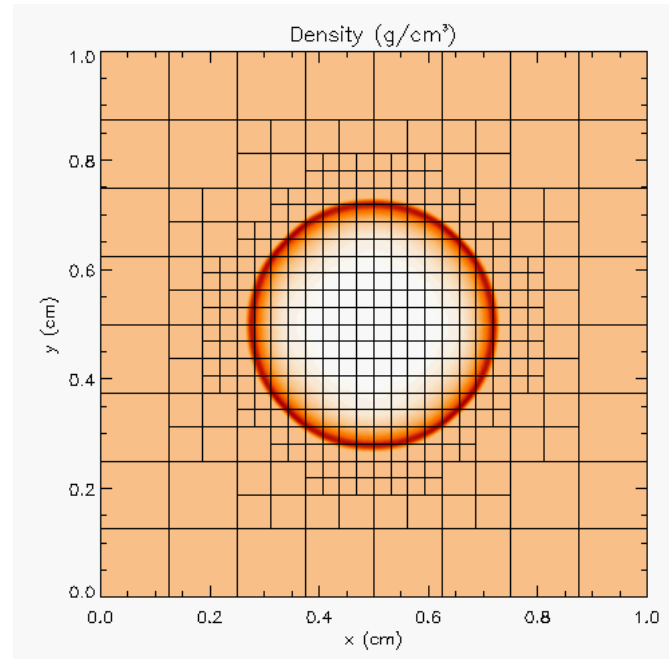
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Strategy for development  
Think of an application with  
analytical solution



- Sedov blast wave
- High pressure at the center
- Shock moves out in a circle
- Analytical solution for how far the shock has travelled

# Step 1 – Equation of State

- Initialize density and internal energy with known values
- Compute pressure and temperature using EOS
- Next use density and computed pressure as input and compute internal energy and temperature using EOS
- Compare computed values against initialized values

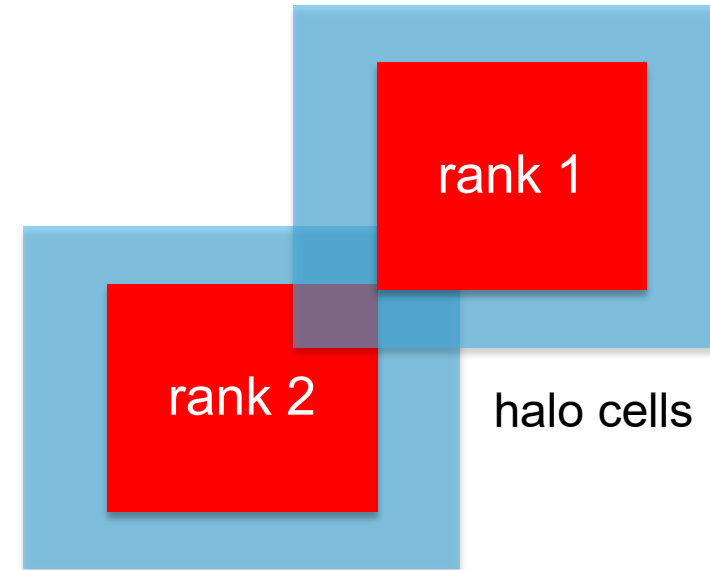
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We have a unit test

## Step 2 – Mesh

- Start with uniform grid
- Domain decomposition for parallelization
  - Halo fill operation
- Initialize the interior (red) with a known function
- Apply halo fill
- Compute values for the halo using the known function
- Compare against filled values





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- Compute values for the halo using the known function
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We have another unit test

## Step 3 – Hydrodynamics

- Apply initial conditions to the mesh
  - zeroes everywhere except at the center
- Write code for the analytical expression of the distance traveled by the shock
- Do time integration
- At time  $T$  compare evolved solution against analytical solution

If both mesh and EOS unit test pass, then any failure is in Hydrodynamics  
This is a composite unit test

This is also the idea behind scaffolding

## Step 4: AMR

- The same halo fill unit test for mesh also works for AMR
- Additional functionalities to test are:
  - Fine-coarse boundary resolution
  - Regridding
- Steps in testing
  - Run Sedov with UG
  - Run Sedov with AMR, but no dynamic refinement
    - If failed fault is in flux correction
  - Run Sedov with AMR and dynamic refinement
    - If failed fault is in regridding

## Step 4: AMR

- The same halo fill unit test for mesh also works for AMR
- Additional functionalities to test are:
  - Fine-coarse boundary resolution
  - Regridding
- Steps in testing
  - Run Sedov with UG
  - Run Sedov with AMR, but no dynamic refinement
    - If failed fault is in flux correction
  - Run Sedov with AMR and dynamic refinement
    - If failed fault is in regridding

We have continued to build scaffolding and are using granular testing to pinpoint the cause of error

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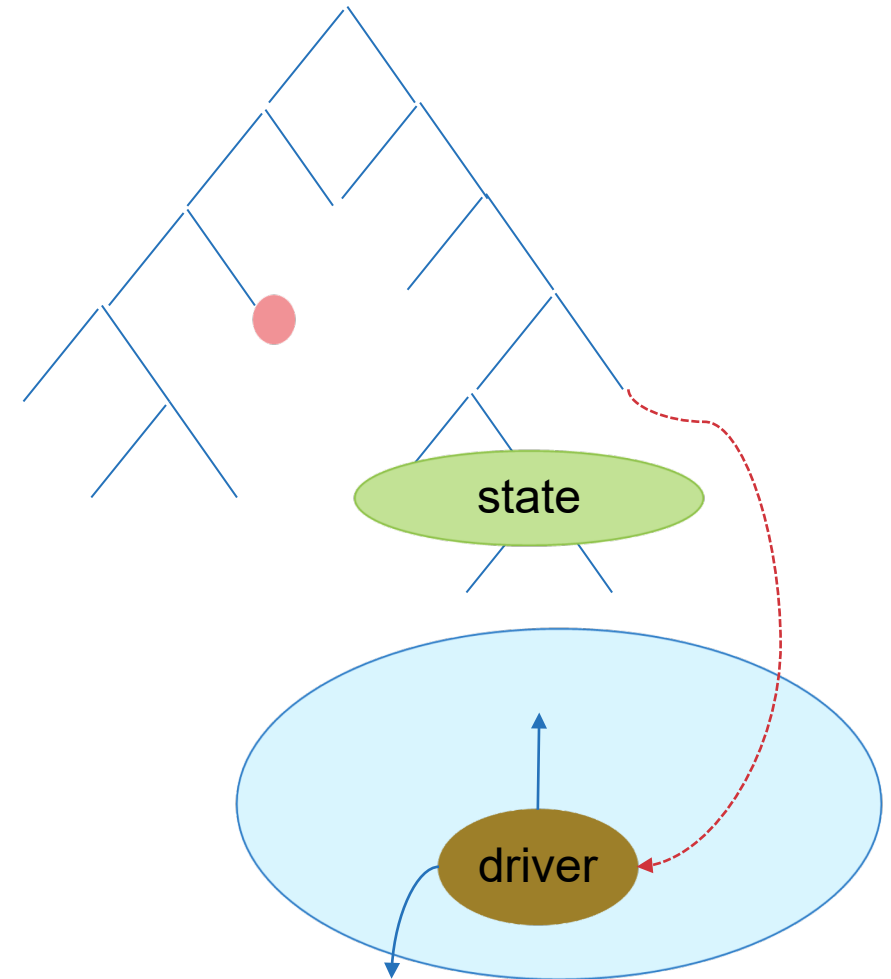
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All of these are examples of white box testing

# Mixed White/Black Box Testing For a Legacy Code

There may not be existing tests

- Isolate a small area of the code
- Dump a useful state snapshot
- Build a test driver
  - Start with only the files in the area
  - Link in dependencies
    - Copy if any customizations needed
- Read in the state snapshot
- Restart from the saved state
- Verify correctness
  - Always inject errors to verify that the test is working



# How to build your test suite?

- A mix of different granularities works well
  - Unit tests for isolating component or sub-component level faults
  - Integration tests with simple to complex configuration and system level
  - Restart tests
- Rules of thumb
  - Simple
  - Enable quick pin-pointing

Useful resources <https://bssw.io/items?topic=testing>

# How do we determine what tests are needed?

## Code coverage tools

- Expose parts of the code that aren't being tested
  - gcov - standard utility with the GNU compiler collection suite (we will use it in the next few slides)
  - Compile/link with `-coverage` & turn off optimization
  - Counts the number of times each statement is executed
  - Necessary for testing, but not sufficient
- gcov also works for C and Fortran
  - Other tools exist for other languages
  - Jcov for Java
  - Coverage.py for python
  - Devel::Cover for perl
  - profile for MATLAB
- Lcov
  - a graphical front-end for gcov
  - available at <https://github.com/linux-test-project/lcov>
  - Codecov.io in CI module
- Hosted servers (e.g., coveralls, codecov)
- graphical visualization of results
- push results to server through continuous integration server



# Good Rules of Thumb

- Test your tests!
  - Make sure tests fail when they're supposed to!
- Add “regression tests”
  - Ensure that bugs aren't creeping in
- Test regularly
  - Critical when teams are adding code regularly
  - To identify and document where changes to the underlying platform change code behavior/results
- Automate regular testing
  - Inculcate the discipline of monitoring the outcome of regular testing
- Exercise third-party dependencies
- Physics/math-based strategies
  - Conserved quantities, symmetries, synthetic operators
  - Eliminate complete dependence on bitwise reproducibility

# Summary

- A testing strategy is essential for producing reliable trustworthy software
  - Invest the time needed to thoroughly test your software at all levels
  - Use automation whenever possible
- Different challenges are associated with exploratory, legacy, and composable codes
  - Adapt your strategy to fit your situation.
  - Eventually you will want to be able to verify all components in a code release.
- Don't get distracted by all the technologies out there – focus on exercising your code.
  - Scaffolding projects can help with mechanics.

# Resources

- Oberkamp, W., & Roy, C. (2010). Verification and Validation in Scientific Computing. Cambridge: Cambridge University Press.  
doi:[10.1017/CBO9780511760396](https://doi.org/10.1017/CBO9780511760396)
- Michael Feathers. 2004. Working Effectively with Legacy Code. Prentice Hall PTR, USA. ISBN: [9780131177055](https://www.isbn-international.org/product/9780131177055)
- A Dubey, K Weide, D Lee, J Bachan, C Daley, S Olofin... - Ongoing Verification of a Multiphysics Community Code. Software: Practice and Experience, 2015  
<https://doi.org/10.1002/spe.2220>