

# It work's on my machine

A workshop on software testing

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# Menu for today

- Basics of testing (theory) and a few buzzwords
- Introduction to `pytest` and `Catch2` frameworks
- Some bits about organizing code / packaging
- Some pointers for setting up CI
- Exercises (and food for thought)

# What not to expect

- Introduction to c++ / python, git or cmake
  - See [the HSF Training Courses](#) for that
- An indepth lecture about all things to know about testing
- A complete guide on pytest or Catch2
  - Check the docs [pytest](#), [Catch2](#)
- A complete guide on setting up CI
- A complete guide on packaging
- An overview of available unittest frameworks

# Why write tests at all?

- “This worked on my machine!”
- “Ah yeah, this issue, we fixed it a few months ago. I wonder why it’s not working now?”
- “I only changed this one line, why is everything broken now?”
- “Why is debugging so hard? Isn’t there an easier way?”

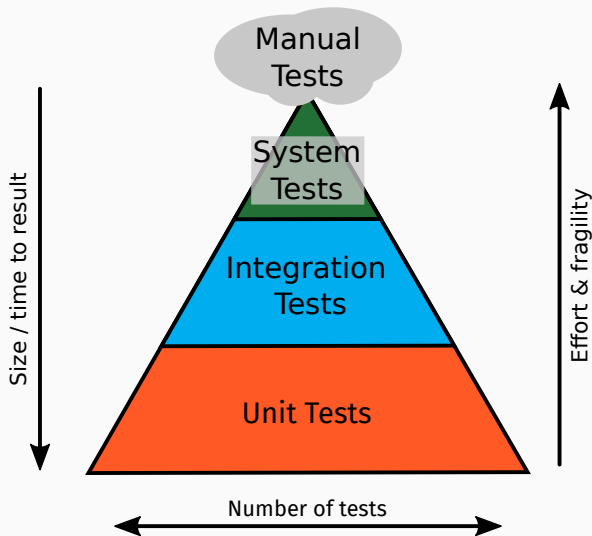
# What tests can do for you

- Confidence that things work (and keep working)
  - Enabling refactoring without breaking things
- *Shift left* in the debugging process
  - Everything that has a **meaningful test** can be excluded from debugging
- More modular code
- Builtin (and up to date) examples and documentation for your code
- Nice status badges on your repository (if you run them in CI)

## What it (unfortunately) can't do

- Make sure there are no bugs in your code at all

# The testing pyramid



- Focus on unit tests today
- Only unit tests is not enough for a working system
- Only integration / system tests are hard to maintain and interpret
- **Unit tests are**
  - quick to run
  - small in scope
  - independent of each other
  - large in numbers

# My first unittest

## Code to test

```
def add_one(number):  
    return number + 1
```

```
int add_one(int number) {  
    return number + 1;  
}
```

## Unit test

```
def test_add_one():  
    assert basics.add_one(3) == 4
```

```
TEST_CASE("add_one") {  
    REQUIRE(basics::add_one(3) == 4);  
}
```

python (pytest)

C++ (Catch2)

# Anatomy of a unit test

- **(Arrange)**
  - Setup things that you want to test (if necessary)
- **Act**
  - Run the function that you want to test
- **Assert**
  - Check that the results of the function are as expected
- **(Cleanup)**
  - Cleanup resources (if necessary)



# Writing testable code

```
def monster_function(fn, options):  
    # ----- collect data -----  
    # ... (20 lines of code)  
    # ... I mean it just takes a bit to do that  
    # ...  
  
    # ----- calculate stuff -----  
    # ... (50 lines of code)  
    # ... Obviously these are non-trivial calculations  
    # ... They require many comments to explain as well  
    # ... So many lines of code  
    # ...  
  
    # ----- make plots -----  
    # ... (50 lines of code)  
    # ... We could do this in 10 lines maybe?  
    # ... But we want nice plots  
    # ... We just have to write this once  
    # ... Then we simply copy it to the next function  
    # ... I wonder if there is a better way to do this  
    # ...  
  
    # ----- write output file -----  
    # ... (20 lines of code)  
    # ... Otherwise what's the point?  
    # ...
```

# Writing testable code

- Small functions / classes
- Functions / classes with a single purpose
- “How can I test this?”

```
def collect_data(fn):  
    # collects data  
    return data  
  
def calc_stuff(data, options):  
    # calculate result  
    return result  
  
def make_plots(data, result):  
    # make plots  
    return plots  
  
def write_output(plots, fn):  
    # write the output  
  
def monster_function(fn, options):  
    data = collect_data(fn)  
    res = calc_stuff(data, options)  
    plots = make_plots(data, res)  
    write_output(plots, fn + ".out")
```

# Writing testable code

- Small functions / classes
- Functions / classes with a single purpose
- “How can I test this?”
- Side benefits:
  - Reusable code
  - Easier to maintain
  - Easier to understand

```
def collect_data(fn):  
    # collects data  
    return data  
  
def calc_stuff(data, options):  
    # calculate result  
    return result  
  
def make_plots(data, result):  
    # make plots  
    return plots  
  
def write_output(plots, fn):  
    # write the output  
  
def monster_function(fn, options):  
    data = collect_data(fn)  
    res = calc_stuff(data, options)  
    plots = make_plots(data, res)  
    write_output(plots, fn + ".out")
```

# Practical considerations for unit tests

- Also test the *error path*, not just the *happy path*

```
def foo(a):  
    if not isinstance(a, int):  
        raise ValueError()  
    return a * 2
```

```
def test_foo_invalid_input():  
    with pytest.raises(ValueError):  
        foo(3.14)
```

python (pytest)

```
int bar(int a) {  
    if (a == 0) {  
        throw std::runtime_error("");  
    }  
    return 42 / a;  
}
```

```
TEST_CASE("bar invalid input") {  
    REQUIRE_THROWS_AS(bar(0),  
        std::runtime_error);  
}
```

C++ (Catch2)

# Practical considerations for unit tests

- Floating point comparisons are *hard* (see e.g. [python docs](#))

```
from pytest import approx

def test_floats():
    # This will fail!
    assert 0.1 + 0.2 == 0.3
    # This will pass
    assert 0.1 + 0.2 == approx(0.3)
```

python (pytest)

```
using namespace Catch::Matchers;

TEST_CASE("float comparison") {
    // This will fail
    REQUIRE(0.1 + 0.2 == 0.3);
    // This will pass
    REQUIRE_THAT(0.1 + 0.2,
                 WithinAbs(0.3, .1));
}
```

C++ (Catch2)

- Need to choose required / desired accuracy up-front!

# Practical considerations for unit tests

- *Test fixtures* allow you to create inputs for tests
  - Handy if the setup requires a few steps
  - Useful when tests require resources (db connection, some running server, ...)
- Unit tests should be **very** easy to read and understand
- Using random inputs for unit tests is possible but the assertions are hard to get right
  - Even statistical tests fail from time to time (by design)
  - It can be an extremely good way of finding bugs ([Fuzzing](#))
- A *flaky* test quickly becomes useless
- Ignored tests are (almost) worse than no tests

# General considerations for tests

- Try to group tests
  - Allows you to run only “interesting” tests during development
- Turn issues into test cases
  - Invest the time to create a minimal reproducer
  - *Red-Green Testing*
- Almost anything that can be automated / scripted can be turned into a (integration / system) test
  - (Usually) trivial to integrate into CI
- Some things don't need tests

# Continuous Integration - CI

- Frequently integrate code changes into main branch
- Quick feedback loop
- Often synonymous with “having tests that run automatically on repository”



[Correct command for installing requirements](#)

Thomas Madlener authored 1 day ago



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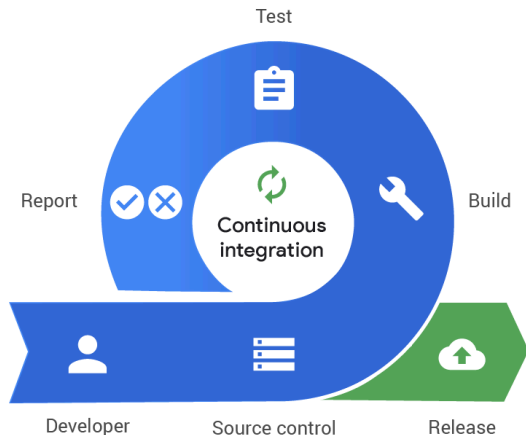


[Make sure to use correct env in github CI](#)

Thomas Madlener authored 1 day ago



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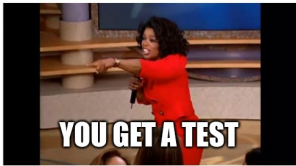
# CI - Some technicalities

- Available on gitlab ([GitLab CI/CD](#)) and github ([GitHub Actions](#))
- Very similar working principles
  - YAML as config language (but different grammars)
- See [our introduction](#) for the basics of GitLab CI/CD
- Exercises have configuration for both
  - Push to github if you want to see github actions
  - Slightly different approaches for dependencies

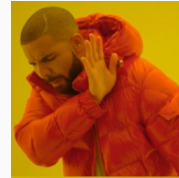
# Considerations for CI

- You can run more stuff than just tests in CI
  - E.g. static analyzers, linters, documentation generation & deployment
- Keep the feedback loop as short as possible
  - Consider stages for running quick tests first
- Run on different platforms / compilers, ...
- Keep dependencies stable
- Include enough output to diagnose the problem quickly
  - Consider storing artifacts that can be inspected
- Make it possible to reproduce the CI environment locally

# Summary



- (Non-trivial) software requires tests
- Unit tests form the basis
- Test automation enables CI
- Even tested software will have bugs



Manually testing software



Running automated tests in CI



# Exercises

- There are two sets of prepared exercises for either python or c++
- Very similar examples to start with, some obvious differences in setup
- The intermediate / advanced examples differ between the two
- Solutions included, you can also do them on your own after the workshop
- Basic configuration for CI already in place
- **Pick one to start with for now**
- **Initial setup interactively together in first half / now**
- Second half of workshop to work on exercises on your own
  - Call us for help whenever necessary

 [.desy.de/fh-sustainability-forum/sustainable-coding-tutorial/python-unittesting](https://github.com/fh-sustainability-forum/sustainable-coding-tutorial/python-unittesting)

 [.desy.de/fh-sustainability-forum/sustainable-coding-tutorial/cpp-unittesting](https://github.com/fh-sustainability-forum/sustainable-coding-tutorial/cpp-unittesting)