



Writing clean scientific software

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Many of these tips come from resources such as: Clean Code and Clean Architecture by R. Martin, The Pragmatic Programmer by Thomas & Hunt, Design Patterns by Gamma et al., and Best Practices for Scientific Computing by G. Wilson et al.

Where I'm coming from...

- These suggestions do not come from:
 - Years of experience writing clean code
- Rather, these suggestions come from:
 - Years of experience writing messy code
 - And then living with the consequences...



Common pain points with scientific software

- Lack of user-friendliness
- Difficult installation
- Inadequate documentation
- Unreadable code
- Cryptic error messages
- Missing tests
- Often not openly available

Why do these pain points exist?

- Programming not covered in science courses
- Scientists tend to be self-taught programmers
- Worth often measured by number of publications
- Code is often written in a rush
- Time pressure prevents us from taking time to learn
- Software not valued as a research product

Publication-driven development (PDD)

- Measure worth of researchers by number of publications
- Write code in a rush to get articles published
- Deprioritize user-friendliness
- Prioritize journal articles over documentation & tests
- Fund research projects, not infrastructure & maintenance
- Avoid training and hiring <u>research software engineers</u>
- Build up <u>technical debt</u> over time

PDD gives us legacy code!

Alternative: Sustainability-driven development

- Cover research software engineering in coursework
- Grow open source software ecosystems
- Invest in long-term health of research software
- Regularly refactor code to reduce technical debt
- Prioritize documentation & continuous integration testing
- Shift towards executable research articles
- Develop code as a community
- Value clean code

My definition of clean code

- Readable
- Easy to change
- Communicates intent
- Well-tested
- Well-documented
- Succinct
- Navigable
- Lets us understand the big picture and little details
- Makes research fun!

"Code is communication!"

Which is more readable?

```
>>> omega_ce=1.76e7*B
>>> electron_gyrofrequency = e * B / m_e
```

How do we choose good variable names?

- Reveal intention and meaning
- Avoid ambiguity
 - Is electron_gyrofrequency an angular frequency?
 - Is volume in cm³ or in barn-megaparsecs? ••
- Be consistent
 - Use one word for each concept
- Use searchable and pronounceable names
- Choose clarity over brevity
 - Longer names are better than unclear abbreviations

Measure the length of a variable name not by the number of characters, but by the time needed to understand its meaning!

Change numbers to named constants

• In this expression:

```
velocity = -9.81 * time
```

- Where does -9.81 come from?
- Are we sure it's correct?
- What if we go to a different planet?
- Use named constants to clarify intent:

```
velocity = gravitational_acceleration * time
```

Use quantities with units instead of numbers

• In this expression:

```
velocity = -9.81 * time
```

- What units does -9.81 have?
- Use a units package to prevent \$328M mistakes



```
from astropy import units
acceleration = -9.81 \times \text{units.meter} / \text{units.second**2}
time = 15 * units.second
velocity = acceleration * time
```

Decompose large programs into functions

- Huge chunks of code are hard to:
 - Read
 - Test
 - Keep track of in our mind
- Breaking code into functions helps us:
 - Reuse code
 - Improve readability
 - Improve testability
 - Isolate bugs

Don't repeat yourself (DRY)

- Copying and pasting code is fraught with peril
 - Bugs would need to be fixed for every copy
- Create functions instead of copying code
 - Simplifies fixing bugs
 - Reduces code duplication
- To change one thing in the code, we should only need to change it in one place

How do we write clean functions?

- Functions should:
 - Be short
 - Do one thing
 - Have no side effects
- Use pure functions

Complex control flow makes code hard to read

```
def is_electron(charge, mass):
    if isclose(charge, -1.67e-19):
        if isclose(mass, 9.11e-31):
            return True
    else:
        return False
```

- Nested if/else statements and for loops make code:
 - Harder to understand
 - Harder to modify
 - More bug-prone

Use guard clauses instead of nested conditionals

```
def is_electron(charge, mass):
    if not isclose(charge, -1.67e-19):
        return False
    if not isclose(mass, 9.11e-31):
        return False
    return True
```

• Take care of edge cases first to simplify subsequent code

Document each function

- State what the function does
- Describe arguments provided to the function
- Describe the value returned by the function
- Include usage examples
- Include additional notes & references as necessary

Questions?

High-level vs. low-level code

High-level code

- Describes the big picture
- Abstracts away implementation details

Low-level code

- Describes implementation details
- Contains concrete instructions for a computer

High-level vs. low-level cooking instructions

- High-level: describe goal of recipe
 - Bake a cake
- Low-level: a line in a recipe
 - Add 1 barn-Mpc of baking powder to flour

Avoid mixing low-level & high-level code

- Mixing low-level & high-level code makes it harder to:
 - Understand what the program is doing
 - Change the implementation
- Separate high-level, big picture code from low-level implementation details

Write code as a top-down narrative*

To perform a numerical simulation, we:

- 1. Read in the inputs
- 2. Set initial conditions
- 3. Perform the time advances
- 4. Output the results

^{*}This is called the Stepdown Rule in Clean Code by R. Martin.

Write code as a top-down narrative

To perform a numerical simulation, we:

- 1. To **read in the inputs**, we:
 - 1.1. Open the input file
 - 1.2 Read in each individual parameter
 - 1.3. Close the input file
- 2. Set initial conditions
- 3. Perform the time advances
- 4. Output the results

Write code as a top-down narrative

To perform a numerical simulation, we:

- 1. To read in the inputs, we:
 - 1.1. Open the input file
 - 1.2. To read in each individual parameter, we:
 - 1.2.1. Read in a line of text
 - 1.2.2. Parse the text
 - 1.2.3. Store the variable
 - 1.3. Close the input file
- 2. Set initial conditions
- 3. Perform the time advances
- 4. Output the results

How do we apply this stepdown rule?

```
def calibrate_observation(raw_image):
    # Subtract bias
    (~20 lines of code)
    # Remove dark current
    (~20 lines of code)
    # Flag cosmic rays
    (~20 lines of code)
```

- This function does more than one thing!
- What if we want to do only one of these steps?
- How do we test each individual step?

The extract function refactoring pattern

Convert each section of code into its own function:

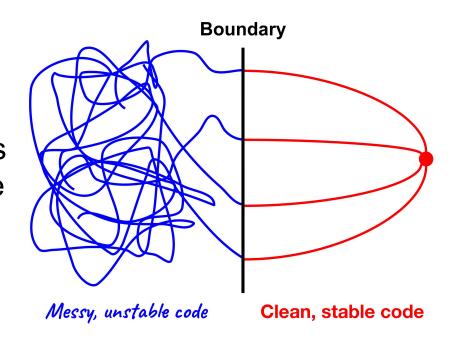
```
def subtract_bias(image): ...
def remove_dark_current(image): ...
def flag_cosmic_rays(image): ...
def calibrate_observation(raw_image):
    image_level1 = subtract_bias(raw_image)
   image_level2 = remove_dark_current(image_level1)
   image_level3 = flag_cosmic_rays(image_level2)
    return image_level3
```

"Program to an interface, not an implementation"

- Suppose our program uses atomic data
- We're using the Chianti database, but want to use AtomDB
- If our high-level code repeatedly calls Chianti, then...
 - Switching to AtomDB will be a pain!
- If our high-level code calls functions that call Chianti...
 - We need only make these interface functions call AtomDB instead
 - The high-level code can remain unchanged!

These interface functions represent a boundary

- Put a **boundary** between stable & unstable code
- The clean, stable code depends directly on the boundary, not the messy unstable code
- The boundary should be stable



Strive for high cohesion & low coupling

- Cohesion is the degree to which the contents of a module belong together
- Coupling is the degree to which the contents of a module depend on other modules
- Code elements that change together at the same time for the same reasons belong together
- Separate code elements that do not change with each other

Comments are not inherently good!

- As code evolves, comments often:
 - Become out-of-date
 - Contain misleading information
 - Get displaced from the corresponding code
- "A comment is a lie waiting to happen"



Potentially unhelpful comments

Commented out code

- Quickly becomes irrelevant
- Keep track of old code using <u>version control</u> instead

Definitions of variables

Encode definitions in variable names instead

```
# torque ← definition in comment
tau = ...

torque = ... ← definition given in variable name
```

Redundant comments

```
i = i + 1 # increment i
```

Helpful commenting practices

- Prefer refactoring code over explaining how it works
- Explain the intent and interface
- Amplify important points
- Explain why an approach was not used
- Provide context and references
- Explain concepts unfamiliar to readers
- Update comments when updating code

Avoid premature optimization of code

- Readability is usually more important than speed
 - Computers are fast and getting faster
 - Our time is more valuable than computing time
- A fold improvement is irrelevant for code that takes a millisecond to run and is only run occasionally
- We should optimize code:
 - Only when necessary
 - After the code is working correctly
 - After using a profiler to identify bottlenecks
- But plan ahead when writing numerically intensive code!

When should we write clean code?

- Some clean coding habits save time quickly
 - Writing short functions that do one thing
 - Writing tests that can be run automatically
- We don't need particularly clean code when we're interactively exploring a data set
- Investing extra time is worthwhile if:
 - You'll re-use the code
 - The code will be shared with others
- Avoid perfectionism

Final thoughts

- Think in terms of tradeoffs
- Break up complicated code into manageable chunks
 - Write short functions that do one thing
 - Separate big picture code from implementation details
- Code is communication
 - Remember importance of community
 - Psychological safety is vital
- Download slides for bonus content

Questions?

Bonus content

Error messages are vital documentation

- The best error messages help users pinpoint a problem and understand how to fix it
- Cryptic error messages can cause hours of frustration

How do we write clean error messages?

- Error messages should:
 - State the problem
 - Describe why it happened
 - Help us fix the problem
- Error messages should be:
 - Helpful!
 - Friendly and supportive
 - Concise, but complete
 - Understandable to new users & contributors
- Provide enough information to solve the problem with minimal extraneous information

Well-written tests make code more flexible

- Without tests:
 - Changes might introduce hidden bugs
 - Less likely to change code for fear of breaking something
- With clean tests:
 - We know if a change broke something
 - We can track down bugs more quickly
- "Legacy code is code without tests."

Why do we write tests?

- To catch and fix bugs
 - Preferably as soon as we introduce them
- To provide confidence that our code gives correct results
- To define what "correct" behavior is
- To show future developers how code should be used
- To keep track of bugs to be fixed later
- In preparation for planned features
- So we can change the code with confidence that we are not introducing hidden bugs elsewhere in the program

Unit tests

A unit test:

- Verifies a single unit of behavior,
- Does it quickly, and
- Does it in isolation from other tests.
- Well-written unit tests
 - Increase code reliability
 - Simplify finding & fixing bugs
 - Make code easier to change

A minimal software test

```
def test_addition():
    """Test adding two integers."""
    assert 1 + 1 == 2, "Incorrect value for 1 + 1"
```

- Descriptive name
- Descriptive docstring (if unclear from name)
- An assertion that a condition is met
- Descriptive error message if condition is not met

Common unit test pattern: arrange, act, assert

Testing best practices

- Write readable and maintainable tests
 - Low quality tests cause future frustrations
- Write tests while writing the code being tested
 - A test delayed is usually a test not written
- Automate tests
 - Make sure tests can be run with ≤ 1 command
- Run tests often!!!!
 - Change 1 thing & run tests ⇒ easier to isolate location of bugs
 - Change 37 things & run tests ⇒ hard to find location of bugs

Testing best practices

Keep tests small

- Avoid multiple assertions per test (unless closely related)
- Avoid conditionals & complex test logic

Keep tests fast

If necessary, add an option to skip slow tests

Keep tests independent of each other

Interdependent tests are harder to change

Make tests deterministic

- Hard to tell when a test that fails intermittently is fixed
- Specify the random seed

Testing best practices

Avoid testing implementation details

Tests of implementation details make code harder to refactor

Turn every bug into a new test

- Helps us fix a bug and prevent it from happening again
- Bugs happen in clusters consider adding related tests

Use a <u>code coverage</u> tool

- Tells us which lines are covered by a test and which are not
- Helps us write targeted tests and find unused code

Consider refactoring code that is difficult to test

Write short functions that do one thing with no side effects

Test-driven development

- More common practice:
 - Write a function
 - Write tests for that function
 - Fix bugs in the function
- <u>Test-driven development</u>
 - Write a failing test
 - Write code to make the test pass
 - Clean up code after tests are passing
- Advantages of writing tests first
 - Makes us think about what each function will do
 - Saves us time
 - Reduces frustration

How do we know what tests to write?

- Test some typical cases
- Test special cases
 - If a function acts weird near 0, test at 0
- Test at and near the boundaries
 - If a function requires a value ≥ 1, test at 1 and 1.001
- Test that code fails correctly
 - If a function requires a value ≥ 1, test at 0.999

Test known solutions and properties

- Test against exact solutions
 - Waves, etc.
- Test equilibrium configurations
- Test against conservation properties
 - Conservation of mass, momentum, & energy
- Test <u>convergence</u> properties
 - Example: test that a 4th order accurate numerical algorithm actually is 4th order
- Test limiting cases

The nascent field of research software engineering

- Research software engineers (RSEs) include
 - Researchers who spend most of their time programming
 - Software engineers developing scientific software
 - Everyone in between
- Challenges
 - Unclear career paths for RSEs
 - Insufficient training for scientists to become RSEs

Additional suggestions

- Learn <u>version control</u> (e.g., <u>qit</u> and <u>GitHub</u> or <u>GitLab</u>)
- Learn an <u>IDE</u> like <u>Visual Studio</u> or <u>PyCharm</u>
- <u>Refactor</u> code periodically
- Set aside time to learn
- Remember the importance of community
 - A software project is not just code it's people too
 - <u>Psychological safety</u> is vital

Psychological safety references: <u>The Fearless Organization</u> by A. Edmondson; and <u>Beyond Buzzwords and Bystanders: A Framework for Systematically Developing a Diverse, Mission Ready, and <u>Innovative Coast Guard Work- force</u> by K. Young-McLear et al.</u>