

Test Driven Development of Scientific Models

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Outline



1 Introduction

2 Testing

- 3 Testing Frameworks
- 4 Test-Driven Development
- 5 TDD and Scientific/Technical Software
- 6 Example



The Tightrope Act



Software development should not feel like this



The Tightrope Act



... or even like this



The Tightrope Act



Hopefully something more like this



































Natural Time Scales

- Design
- Edit source
- Compilation
- Batch waiting in queue
- Execution
- Analysis









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TDD - Introduction - NCAR

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Some observations



- Risk grows with magnitude of implementation step
- Magnitude of implementation step grows with cost of verification/validation

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- Magnitude of implementation step grows with cost of verification/validation

Conclusion: Optimize productivity by reducing cost of verification!

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Testing





Collection of tests that constrain system





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• Detects unintended changes

Collection of tests that constrain system





Localizes defects



Collection of tests that constrain system



- Detects unintended changes
- Localizes defects
- Improves developer confidence



Collection of tests that constrain system



- Detects unintended changes
- Localizes defects
- Improves developer confidence
- Decreases risk from change







"The main thing that distinguishes legacy code from non-legacy code is tests, or rather a lack of tests."

Michael Feathers Working Effectively with Legacy Code





"The main thing that distinguishes legacy code from non-legacy code is tests, or rather a lack of tests."

Robert E. Marter Serve WORKING LEGACY CODE Michael C. Feathers

Michael Feathers Working Effectively with Legacy Code

Lack of tests leads to fear of introducing subtle bugs and/or changing things inadvertently.

• Programming on a tightrope



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Lack of tests leads to fear of introducing subtle bugs and/or changing things inadvertently.

• Programming on a tightrope

This is also a barrier to involving pure software engineers in the development of our models.







• Takes too much time to write tests







- Takes too much time to write tests
- Too difficult to maintain tests





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- It takes too long to run the tests





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http://java.dzone.com/articles/unit-test-excuses - James Sugrue

Just what is a test anyway?

Tests can exist in many forms

• Conditional termination:

```
IF (PA(I,J)+PTOP.GT.1200.) &
call stop_model('ADVECM: Pressure diagnostic error',11)
```

Diagnostic print statement

print *, 'loss of mass = ', deltaMass

• Visualization of output





Analogy with Scientific Method?



Reality Constraints: theory and data Formulate hypothesis Perform experiment Refine hypothesis

- Requirements
- Constraints: tests
- Trial implementation
- \longrightarrow Run tests

 \longrightarrow

 \longrightarrow

 \longrightarrow Refine implementation

Properties of good tests




- Isolating
 - Test failure indicates location in source code



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- Orthogonal
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- Complete
 - Each bit of functionality covered by at least one test

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 - No side effects
 - Test order does not matter
 - Corollary: cannot terminate execution







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- Clear intent









testTrajectory() ! $s = \frac{1}{2}at^2$

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$$a = 2.; t = 3.$$





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$$\mathsf{s} = \mathsf{trajectory}(\mathsf{a}, \mathsf{t})$$

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testTrajectory() ! $s = \frac{1}{2}at^2$

$$a = 2.; t = 3.$$

$$\mathsf{s} = \mathsf{trajectory}(\mathsf{a}, \mathsf{t})$$

call assertEqual (9., s)





testTrajectory() ! $s = \frac{1}{2}at^2$ a = 2.; t = 3.

- s = trajectory(a, t)
- call assertEqual (9., s)

! no op





testTrajectory() ! $s = \frac{1}{2}at^2$

call assertEqual (9., trajectory (2.,3.))

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Testing Frameworks



- Provide infrastructure to radically simplify:
 - Creating test routines (Test cases)
 - Running collections of tests (Test suites)
 - Summarizing results
- Key feature is collection of assert methods
 - Used to express expected results

```
call assertEqual(120, factorial(5))
```

- Generally specific to programming language (xUnit)
 - Java (JUnit)
 - Pnython (pyUnit)
 - C++ (cxxUnit, cppUnit)
 - Fortran (FRUIT, FUNIT, pFUnit)

GUI - JUnit in Eclipse







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🕖 pFUnit

(Somewhat) New Paradigm: TDD



Old paradigm:

- Tests written by separate team (black box testing)
- Tests written after implementation

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NASA

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Consequences:

- Testing schedule compressed for release
- Defects detected late in development (\$\$)

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New paradigm

- Developers write the tests (white box testing)
- Tests written before production code
- Enabled by emergence of strong unit testing frameworks

The TDD cycle









• High reliability



- High reliability
- Excellent test coverage



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- Excellent test coverage
- Always "ready-to-ship"



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- Always "ready-to-ship"
- Tests act as maintainable documentation
 - Test shows real use case scenario
 - Test is maintained through TDD process



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• Quality implementation?



- Many professional SEs are initially skeptical
 - High percentage refuse to go back to the old way after only a few days of exposure.
- Some projects drop bug tracking as unnecessary
- Often difficult to sell to management
 - "What? More lines of code?"

Not a panacea





• Requires training, practice, and discipline


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- Need strong tools (framework + refactoring)



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- Maintaining tests difficult during a major re-engineering effort.
 - But isnt the alternative is even worse?!!

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pFUnit



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Software Testing vs Science/Validation



Software tests should only check implementation.

- Only a subset tests will express external requirements (i.e. implementation independent)
- Other tests will reflect implementation choices
- Use "convenient" input values not realistic values

Consider tests for an ODE integrator implemented with RK4

- A generic test may be for a constant flow field any integrator should get an "exact" answer
- A RK4 specific test may provide an artificial "flow field" that returns the values 1.,2.,3.,4. on subsequent calls *independent* of the coordinates

Test by Layers





Do test

- Proper # of iterations
- Pieces called in correct order
- Passing of data between components

Do NOT test

• Calculations inside components

Much easier to do in practice with objects than with procedures.





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Unfortunately ...

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Unfortunately ...

- Error estimates are seldom available for complex algorithms
- And of those, usually we just have an asymtotic form with unknown leading coefficient!

Numerical tolerance (cont'd)



Numerical tolerance (cont'd)



Numerical tolerance (cont'd)



Observations

machine epsilon is a good estimate for most short arithmetic expressions



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- 2 large errors arise in small expressions in fairly obvious places $(1/\Delta)$



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- **③** larger errors are generally a result of composition of many operations



- machine epsilon is a good estimate for most short arithmetic expressions
- 2 large errors arise in small expressions in fairly obvious places $(1/\Delta)$
- larger errors are generally a result of composition of many operations
 Conclusion: If we write software as a composition of distinct small functions and subroutines, the errors can be reasonably bounded at each stage



- TDD does not directly relate to issues of stability
- If long integration gets incorrect results:



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- TDD does not directly relate to issues of stability
- If long integration gets incorrect results:
 - Software defect: missing test
 - ② Genuine science challenge
- TDD can reduce the frequency at which long integrations are needed/performed

TDD and Lack of Analytic Results



- Keep in mind: "How can you implement it if you cannot say what it should do?"
- Split into pieces often each step has analytic solution
- Choose input values that are convenient

Consider a trivial case:

```
call assertEqual(3.14159265, areaOfCircle(1.))
call assertEqual(6.28..., areaOfCircle(2.))
```

What if instead the areaOfCircle() function accepted 2 arguments: " π " and *r*.

```
call assertEqual(1., areaOfCircle(1., 1.))
call assertEqual(4., areaOfCircle(1., 2.))
call assertEqual(2., areaOfCircle(2., 1.))
```



- Are the tests as complex as the implementation?
- Short answer: **No**



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- Short answer: **No**
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 - Unit tests use specific inputs implementation handles generic case
 - Each layer of algorithm is tested separately
 - Layers of the production code are *coupled* huge complexity
 - Tests are *decoupled* low complexity

TDD and the Legacy Burden



- TDD was created for developing *new* code, and does not directly speak to maintaining legacy code.
- Adding new functionality
 - Avoid wedging new loging directly into existing large procedure
 - Use TDD to develop separate facility for new computation
 - Just call the new procedure from the large legacy procedure
- Refactoring
 - Use unit tests to constrain existing behavior
 - Very difficult for large procedures
 - Try to find small pieces to pull out into new procedures

TDD Best Practices





 $\bullet\,$ Small steps - each iteration $\ll 10$ minutes



- $\bullet\,$ Small steps each iteration $\ll\,10$ minutes
- Small, readable tests



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- $\bullet\,$ Small steps each iteration $\ll\,10$ minutes
- Small, readable tests
- Extremely fast execution 1 ms/test or less
- Ruthless refactoring
- Verify that each test initially fails



- Optimized algorithms may require many steps within a single procedure
- TDD emphasizes small simple procedures
- Such an approach may lead to slow execution
- Solution: Bootstrapping
 - Use initial solution as unit test for optimized solution
 - Maintain *both* implementations

Experience to date



TDD has been used heavily within several projects at NASA

- Mostly for "infrastructure" portions relatively little numerical alg.
- pFUnit
- DYNAMO spectral MHD code on shperical shell
- GTRAJ offline trajectory integration (C++)
- Snowfake virtual snowfakes; Multi-lattice Snowfake

Observations:

- \sim 1:1 ratio of test code to source code
- Works very well for *infrastructure*
- Learning curve
 - 1-2 days for technique
 - Weeks-months to wean old habits
 - Full benefit may require some sophistication

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Linear Interpolation





Potential Tests





• Bracketing: Find *i* such that $x_i \le \hat{x} \le x_{i+1}$

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- Computing node weights:

$$w_a = \frac{x_{i+1} - \hat{x}}{x_{i+1} - x_i}$$
$$w_b = 1 - w_a$$

Potential Tests



- Bracketing: Find *i* such that $x_i \le \hat{x} < x_{i+1}$
- Computing node weights:

$$w_a = \frac{x_{i+1} - \hat{x}}{x_{i+1} - x_i}$$
$$w_b = 1 - w_a$$

• Compute weighted sum: $\hat{y} = w_a f(x_i) + w_b f(x_{i+1})$



Case	Preconditions		Postcondition
	nodes	x	return



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	nodes	х	return
interior	$\{x\} = \{1, 2, 3\}$	$\hat{x} = 1.5$	i = 1



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other interior	${x} = {1, 2, 3}$	$\hat{x} = 2.5$	<i>i</i> = 2



Case	Preconditions		Postcondition
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at node	$\{x\} = \{1, 2, 3\}$	$\hat{x} = 2.0$	i = 2 (?)



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at node	${x} = {1, 2, 3}$	$\hat{x} = 2.0$	i = 2 (?)
at edge	$\{x\} = \{1, 2, 3\}$	$\hat{x} = 1.0$	i = 1 (?)



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at edge	$\{x\} = \{1, 2, 3\}$	$\hat{x} = 1.0$	i = 1 (?)
other edge	$\{x\} = \{1, 2, 3\}$	$\hat{x} = 3.0$	i = 2 (????)



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other edge	${x} = {1, 2, 3}$	$\hat{x} = 3.0$	i = 2 (????)
out-of-bounds	$\{x\} = \{1, 2, 3\}$	$\hat{x} = 1.5$	out-of-bounds error



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	nodes	х	return
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other edge	${x} = {1, 2, 3}$	$\hat{x} = 3.0$	i = 2 (????)
out-of-bounds	$\{x\} = \{1, 2, 3\}$	$\hat{x} = 1.5$	out-of-bounds error
out-of-order	$\{x\} = \{1, 2, 3\}$	$\hat{x} = 1.5$	out-of-order error



- Preconditions: $\{x\} = \{1, 2, 3\}, \hat{x} = 1.5$
- Postcondition: return 1



- Preconditions: $\{x\} = \{1, 2, 3\}, \hat{x} = 1.5$
- Postcondition: return 1

```
subroutine testBracket1()
nodes = [1.,2.,3.]
index = getBracket(nodes, 1.5)
call assertEqual(1, index)
end subroutine
```



- Preconditions: $\{x\} = \{1, 2, 3\}, \hat{x} = 1.5$
- Postcondition: return 1

```
subroutine testBracket1()
call assertEqual(1, getBracket([1.,2.,3.], 1.5))
end subroutine
```



- Preconditions: $\{x\} = \{1, 2, 3\}, \hat{x} = 1.5$
- Postcondition: return 1

```
subroutine testBracket1()
call assertEqual(1, getBracket([1.,2.,3.], 1.5))
end subroutine
```

```
function getBracket(nodes, x) result(index)
    index = 1
end function
```



- Preconditions: $\{x\} = \{1, 2, 3\}, \hat{x} = 2.5$
- Postcondition: return 2

```
subroutine testBracket2()
nodes = [1.,2.,3.]
index = getBracket(nodes, 2.5)
call assertEqual(2, index)
end subroutine
```



- Preconditions: $\{x\} = \{1, 2, 3\}, \hat{x} = 2.5$
- Postcondition: return 2

```
subroutine testBracket2()
nodes = [1.,2.,3.]
index = getBracket(nodes, 2.5)
call assertEqual(2, index)
end subroutine
```

```
function getBracket(nodes, x) result(index)
    if (x > nodes(2)) then
        index = 2
    else
        index = 1
    end if
end function
```



- Preconditions: $\{x\} = \{1, 2, 3\}, \hat{x} = 2.5$
- Postcondition: return 2

```
subroutine testBracket2()
nodes = [1.,2.,3.]
index = getBracket(nodes, 2.5)
call assertEqual(2, index)
end subroutine
```

```
function getBracket(nodes, x) result(index)
if (x > nodes(2)) then
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else
index = 1
end if
end function
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Generalize ...



- Preconditions: $\{x\} = \{1, 2, 3\}, \hat{x} = 2.5$
- Postcondition: return 2

```
subroutine testBracket2()
nodes = [1.,2.,3.]
index = getBracket(nodes, 2.5)
call assertEqual(2, index)
end subroutine
```

```
function getBracket(nodes, x) result(index)
do i = 1, size(nodes) 1
    if (nodes(i+1) > x) index = i
    end do
end function
```

Tests for Computing Weights



Case	Preconditions		Postcondition
	interval	x	weights

Tests for Computing Weights



Case	Preconditions		Postcondition
	interval	х	weights
lower bound	[1., 2.]	$\hat{x} = 1.0$	w = [1.0, 0.0]

Tests for Computing Weights



Case	Preconditions		Postcondition
	interval	х	weights
lower bound	[1., 2.]	$\hat{x} = 1.0$	w = [1.0, 0.0]
upper bound	[1., 2.]	$\hat{x} = 1.0$	w = [0.0, 1.0]


Case	Preconditions		Postcondition
	interval	x	weights
lower bound	[1., 2.]	$\hat{x} = 1.0$	w = [1.0, 0.0]
upper bound	[1., 2.]	$\hat{x} = 1.0$	w = [0.0, 1.0]
interior	[1., 2.]	$\hat{x} = 1.5$	w = [0.5, 0.5]



Case	Preconditions		Postcondition
	interval	х	weights
lower bound	[1., 2.]	$\hat{x} = 1.0$	w = [1.0, 0.0]
upper bound	[1., 2.]	$\hat{x} = 1.0$	w = [0.0, 1.0]
interior	[1., 2.]	$\hat{x} = 1.5$	w = [0.5, 0.5]
big interval slope	[1., 3.]	$\hat{x} = 1.5$	w = [0.75, 0.25]



Case	Preconditions		Postcondition
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lower bound	[1., 2.]	$\hat{x} = 1.0$	w = [1.0, 0.0]
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big interval slope	[1., 3.]	$\hat{x} = 1.5$	w = [0.75, 0.25]
degenerate	[1., 1.]	$\hat{x} = 1.0$	degenerate error



Case	Preconditions		Postcondition
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lower bound	[1., 2.]	$\hat{x} = 1.0$	w = [1.0, 0.0]
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interior	[1., 2.]	$\hat{x} = 1.5$	w = [0.5, 0.5]
big interval slope	[1., 3.]	$\hat{x} = 1.5$	w = [0.75, 0.25]
degenerate	[1., 1.]	$\hat{x} = 1.0$	degenerate error
out-of-bounds	[1., 2.]	$\hat{x} = 0.5$	out-of-bounds error

Example: Weights Test 1



- Precondition: $[a, b] = [1., 2.], \hat{x} = 1.0$
- Postcondition: $w = \{1.0, 0.0\}$

```
subroutine testWeight1()
   real :: interval(2), weights(2)
   real :: x
   interval = [1.,2.]
   weights = computeWeights(interval, 1.0)
   call assertEqual([1.0,0.0], weights)
end subroutine testWeight1

real function computeWeights(interval, x) result(weights)
```

```
real, intent(in) :: interval(2)
real, intent(in) :: x
weights = [1.0,0.0]
end function
```

Example: Tying it together



• Precondition:

•
$$\{(x, y)_i\} = \{(1, 1), (2, 1), (4, 1)\}$$

• $\hat{x} = 3$

• Postcondition: $\hat{y} = 1$.

```
subroutine testInterpolateConstantY()
real :: nodes(2,3)
nodes = reshape([[1,1],[2,1],[4,1]], shape=[2,3])
call assertEqual(1.0, interpolate(nodes, 3.0))
end subroutine testInterpolate1
```

```
function interpolate(nodes, x)
    real, intent(in) :: nodes(:,:)
    y = 1
end function interpolate
```

Example: Tying it together



- Precondition:
 - $\{(x, y)_i\} = \{(1, 1), (2, 3), (4, 1)\}$

```
► \hat{x} = 3
```

• Postcondition: $\hat{y} = 2$.

```
subroutine testInterpolate1()
    real :: nodes(2,3)
    nodes = reshape([[1,1],[2,3],[4,1]], shape=[2,3])
    call assertEqual(1.0, interpolate(nodes, 3.0))
end subroutine testInterpolate1
```

```
function interpolate(nodes, x) result(y)
integer :: i
real :: weights(2), xAtEndPoints(2), yAtEndpoints(2)
i = getBracket(nodes(1,:), x)
xAtEndPoints = nodes(1,i) ! used derived type?
yAtEndpoints = nodes(2,i)
weights = computeWeights(nodes(1,[i,i+1]), x)
y = sum(weights * yAtEndpoints)
```

end function interpolate

Outline



1 Introduction



- 3 Testing Frameworks
- 4 Test-Driven Development
- 5 TDD and Scientific/Technical Software
- 6 Example



pFUnit - Fortran Unit testing framework



- Tests written in Fortran
- Supports testing of parallel (MPI) algorithms
- Support for multi-dimensional array assertions
- Written in standard F95 (plus a tiny bit of F2003)
- Developed using TDD

Tutorial in the afternoon sessioon

References



- pFUnit: http://sourceforge.net/projects/pfunit/
- Tutorial materials
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- Test-Driven Development: By Example Kent Beck
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