

Test Driven Development of Scientific Models

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Outline



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

The Tightrope Act



Software development should not feel like this



The Tightrope Act



... or even like this



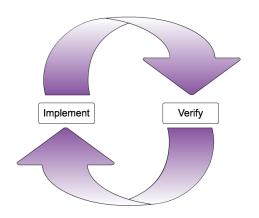
The Tightrope Act



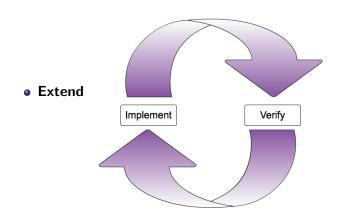
Hopefully something more like this



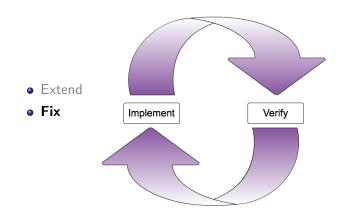




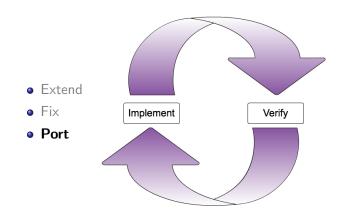




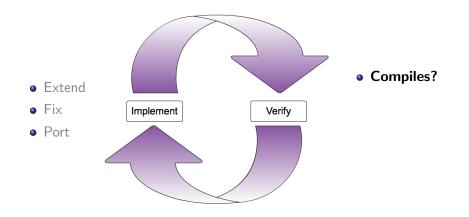




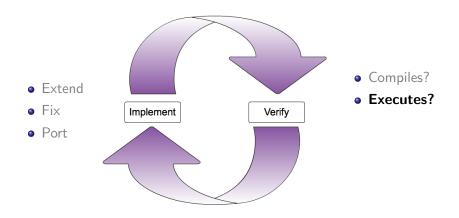




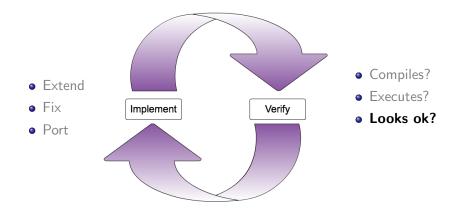




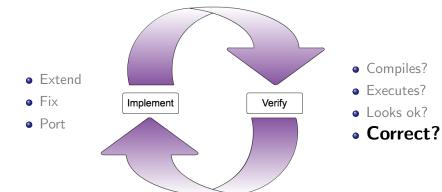












Natural Time Scales

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









Some observations



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

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Conclusion:

Optimize productivity by reducing cost of verification!

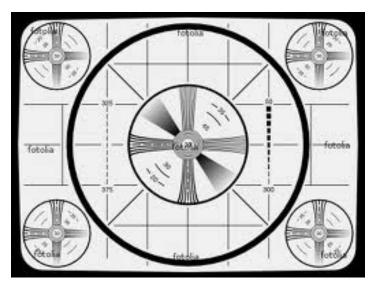
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Testing









NASA

Collection of tests that constrain system



Detects unintended changes

NASA



- Detects unintended changes
- Localizes defects

NASA



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- Improves developer confidence

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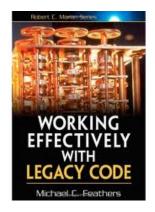
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- Improves developer confidence
- Decreases risk from change





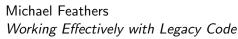
"The main thing that distinguishes legacy code from non-legacy code is tests, or rather a lack of tests."

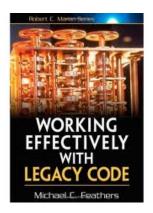
Michael Feathers
Working Effectively with Legacy Code





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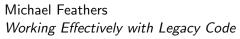


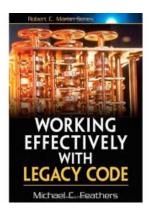
Lack of tests leads to fear of introducing subtle bugs and/or changing things inadvertently.

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





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- Too difficult to maintain tests





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- It takes too long to run the tests





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Just what is a test anyway?



Tests can exist in many forms

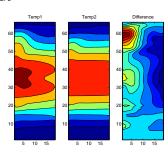
Conditional termination:

```
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

ightarrow Requirements

ightarrow Constraints: tests

ightarrow Trial implementation

 \longrightarrow Run tests

→ Refine implementation

Properties of good tests





- Isolating
 - ▶ Test failure indicates location in source code



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- Orthogonal
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 - Run quickly
 - Small memory, etc.

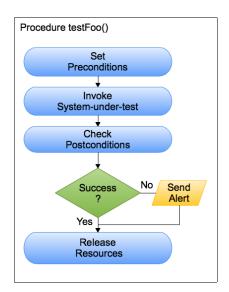


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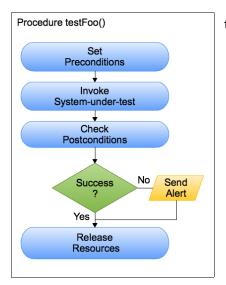
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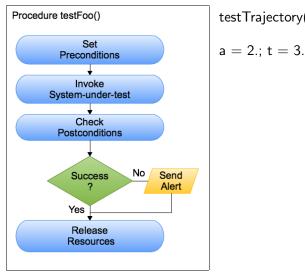
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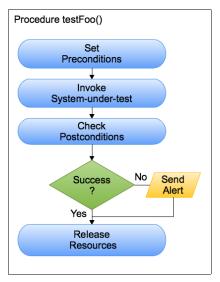
testTrajectory() ! $s = \frac{1}{2}at^2$





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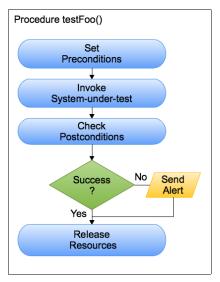


testTrajectory() ! $s = \frac{1}{2}at^2$

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

s = trajectory(a, t)



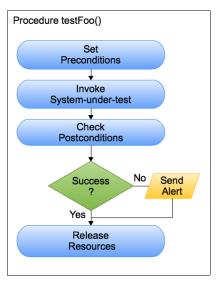


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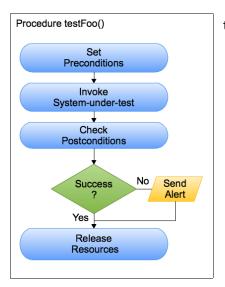
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$$s = trajectory(a, t)$$

! no op





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

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

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Testing Frameworks



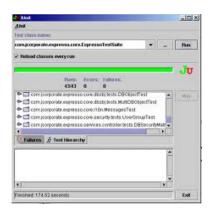
- Provide infrastructure to radically simplify:
 - Creating test routines (Test cases)
 - Running collections of tests (Test suites)
 - ► Summarizing results
- Key feature is collection of assert methods
 - Used to express expected results

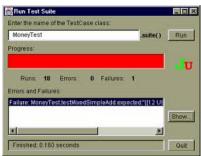
```
call assertEqual(120, factorial(5))
```

- 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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(Somewhat) New Paradigm: TDD



Old paradigm:

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

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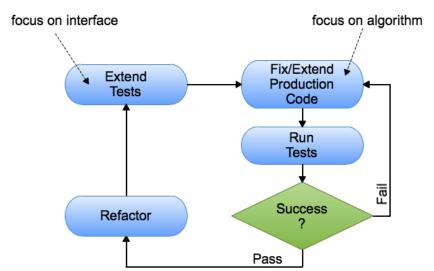
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New paradigm

- 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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- Always "ready-to-ship"



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- Tests act as maintainable documentation
 - Test shows real use case scenario
 - Test is maintained through TDD process



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- Productivity
- Predictable schedule
- Porting
- Quality implementation?

Anecdotal Testimony



- 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



Not a panacea



• Requires training, practice, and discipline



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

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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

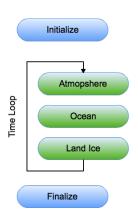


Do test

- 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.





For testing numerical results, a good estimate for the tolerance is necessary:

• If the tolerance is too *low*, then the test may fail for uninteresting reasons.



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 - Error estimates are seldom available for complex algorithms



- If the tolerance is too low, then the test may fail for uninteresting reasons.
- If the tolerance is too high, then the test may have no teeth
 Unfortunately ...
 - Error estimates are seldom available for complex algorithms
 - And of those, usually we just have an asymtotic form with unknown leading coefficient!





Observations



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machine epsilon is a good estimate for most short arithmetic expressions



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- $oldsymbol{0}$ large errors arise in small expressions in fairly obvious places $(1/\Delta)$



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- ② large errors arise in small expressions in fairly obvious places $(1/\Delta)$
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Conclusion: If we write software as a composition of distinct small functions and subroutines, the errors can be reasonably bounded at each stage

TDD and long integration



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- If long integration gets incorrect results:

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TDD and long integration



- TDD does not directly relate to issues of stability
- If long integration gets incorrect results:
 - Software defect: missing test
 - @ 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.))
```



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 - ▶ 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





 \bullet Small steps - each iteration \ll 10 minutes



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



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

TDD and Performance



- 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:

- ullet $\sim 1:1$ 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

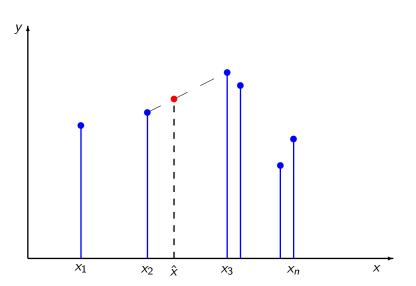
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Linear Interpolation









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



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- Computing node weights:

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

$$w_b = 1 - w_a$$



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$$w_b = 1 - w_a$$

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



Case	Preconditions		Postcondition
	nodes	X	return



Case	Preconditions		Postcondition
	nodes	×	return
interior	${x} = {1, 2, 3}$	$\hat{x} = 1.5$	i = 1



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interior	${x} = {1, 2, 3}$	$\hat{x} = 1.5$	i = 1
other interior	$\{x\} = \{1, 2, 3\}$	$\hat{x} = 2.5$	i=2



Case	Preconditions		Postcondition
	nodes	x	return
interior	${x} = {1, 2, 3}$	$\hat{x} = 1.5$	i = 1
other interior	$\{x\} = \{1, 2, 3\}$	$\hat{x} = 2.5$	i=2
at node	$\{x\} = \{1, 2, 3\}$	$\hat{x} = 2.0$	i=2 (?)



Case	Preconditions		Postcondition
	nodes	×	return
interior	${x} = {1, 2, 3}$		
other interior	${x} = {1, 2, 3}$		
at node	${x} = {1, 2, 3}$	$\hat{x} = 2.0$	i = 2 (?)
at edge	$\{x\} = \{1, 2, 3\}$	$\hat{x} = 1.0$	i=1 (?)



Case	Preconditions		Postcondition
	nodes	×	return
interior	${x} = {1, 2, 3}$	$\hat{x} = 1.5$	i = 1
other interior	${x} = {1, 2, 3}$	$\hat{x} = 2.5$	i=2
at node	$\{x\} = \{1, 2, 3\}$		
at edge	$\{x\} = \{1, 2, 3\}$		
other edge	$\{x\} = \{1, 2, 3\}$	$\hat{x} = 3.0$	i=2 (????)



Case	Preconditions		Postcondition
	nodes	x	return
interior	${x} = {1, 2, 3}$	$\hat{x} = 1.5$	i = 1
other interior	${x} = {1, 2, 3}$	$\hat{x} = 2.5$	i=2
at node	${x} = {1, 2, 3}$	$\hat{x} = 2.0$	i = 2 (?)
at edge	${x} = {1, 2, 3}$	$\hat{x} = 1.0$	$i = 1 \ (?)$
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



Case	Preconditi	Postcondition	
	nodes	x	return
interior	${x} = {1, 2, 3}$	$\hat{x} = 1.5$	i = 1
other interior	${x} = {1, 2, 3}$	$\hat{x} = 2.5$	i=2
at node	${x} = {1, 2, 3}$	$\hat{x} = 2.0$	i = 2 (?)
at edge	${x} = {1, 2, 3}$	$\hat{x} = 1.0$	$i = 1 \ (?)$
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
	1		'



• 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
    index = 2
  else
    index = 1
  end if
end function
```

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
```



Case	Preconditions		Postcondition
	interval	×	weights



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



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



Case	Preconditions		Postcondition
	interval	×	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]
degenerate	[1., 1.]	$\hat{x} = 1.0$	degenerate error



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]
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

subroutine testWeight1()



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

```
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:

```
\begin{cases} \{(x,y)_i\} = \{(1,1),(2,1),(4,1)\} \\ \hat{x} = 3 \end{cases}
```

• 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)

```
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?
   vAtEndpoints = nodes(2,i)
   weights = computeWeights (nodes (1,[i,i+1]), \times)
   y = sum(weights * yAtEndpoints)
end function interpolate
```

nodes = reshape ([[1,1],[2,3],[4,1]], shape = [2,3])

Outline



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

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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