

Employing property-based testing

Some experiences from testing linalgwrap

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Property-based testing

- Testing technique for unit tests
- Auto-generated test cases
- Extremely useful tool to *localise* a bug
- Originally QuickCheck framework in Haskell
- Nowadays widely available

General idea

- Take what we know about our code
- Generate the test cases from that
- *Preconditions:*
 - Requirements before the run
 - ⇒ Data generation / state setup
- *Postconditions:*
 - Guaranteed state after the run
 - ⇒ Assertions to check for

General idea

- Take what we know about our code
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Test case generation

- Use random seed
 - Generate multiple sets of input data
- ⇒ Per run: *Many* different test cases
- Execute from easy to hard
 - On failure: *Shrink*
- ⇒ Try to find simplest failing case

Hierarchy of methods

- Assert postcondition properties:
 - Reversing a string twice gives back the original.
 - Cauchy-Schwarz inequality holds
- Comparative testing:
 - Same behaviour in model and system under test (SUT)
- Random *chain* of operations:
 - Do model and SUT show an equivalent state?

⇒ *Stateful testing*

Software and implementations

- Varying feature sets
- C++: <https://github.com/emil-e/rapidcheck>
- Haskell: <http://www.cse.chalmers.se/~rjmh/QuickCheck>
- Python: <http://hypothesis.works>

Advantages

- Make yourself aware of pre/postconditions
- Different test cases each run
- Test what you did *not* think about
- I.e. fuzz your own program
- True stateful testing possible

Demo

DEMO

`https://github.com/mfherbst/c14h-rapidcheck-catch`

Challenges

- Test cases: Hard enough, but not too hard
 - Complex preconditions:
 - Existence of solution
 - Dimensionality
 - Numerical stability
- ⇒ Separate numerical and implementation error

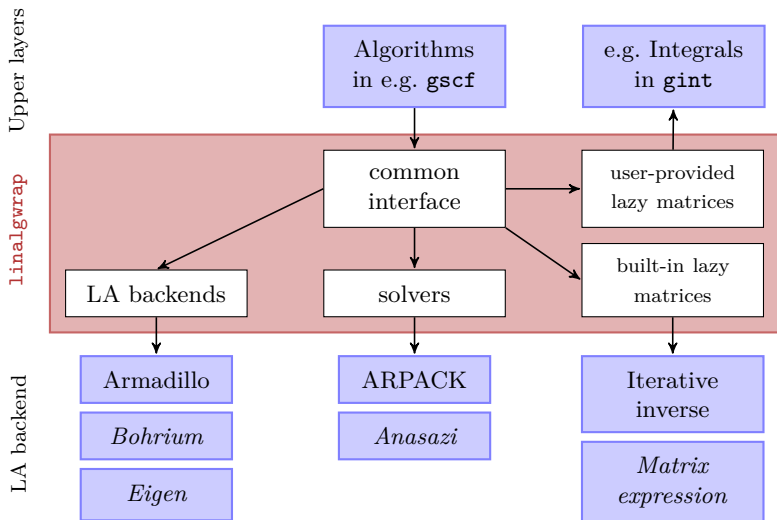
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linalgwrap

Linear algebra wrapper library — <https://linalgwrap.org>



Generators

- Generator for scalars:
 - Values within $[-10^5, -10^{-5}] \cup \{0\} \cup [10^{-5}, 10^5]$
 - Extremes less likely
- Generator for vectors and matrices:
 - The higher the rank the smaller the size
- *Not* checked:
 - Norm
 - Condition number

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Numerics-aware comparison

- Based on `krimis::NumComp`
- Error boundaries relative to *machine epsilon*
- Flexible interface
 - Temporarily tighten / loosen threshold
- Not fast, but informative
- Support for vectors and matrices

Numerics-aware comparison

Example

```
1 // Check that matrices are initialised to zero
2 //
3 using namespace krims;
4 const size_t size = *gen::numeric_size<2>();
5 Matrix m(size,size);
6 Matrix n(size,size);
7 for (auto& elem : n) elem=0.0;
8
9 // Compare and use defaults
10 REQUIRE( numcomp(n) == m );
11
12 // Compare and use specified tolerance
13 REQUIRE( n == numcomp(m).tolerance(1e-13) );
14
15 // Compare and use 0.1 times the default tolerance
16 REQUIRE( n == numcomp(m).tolerance(Lower) );
17
18 // The default is relative to machine epsilon
19 // and can be bumped or decreased locally
```

Vector and matrix operations

- Stored matrices:
 - Generate objects
 - Apply operation
 - Assert equivalence against model
- Lazy matrices:
 - Stateful testing
 - Apply random sequence of operations
 - Compare each time against stored matrix model

Eigenproblems

- No satisfactory input data generation available
- ⇒ Conventional hard-coded test cases
- Not easy to check results against reference
 - E.g. uniqueness up to Unitarian transformation only
- ⇒ Check properties instead:
 - Residual of eigenpairs
 - Size of off-diagonal elements
- Linear problems similar

Possible improvements

- More specific generators:
 - Guarantee certain properties
 - Bounds on the norm or condition number
 - Good matrices for addition/subtraction
 - Spectral properties, e.g. positive definiteness
- Eigenproblems and linear problems:
 - Family of parametrised input problems
 - Well-behaving and solvable
 - Increasing in difficulty

Summary

- Better code:
 - Different mindset when coding
- Better testing:
 - Lower influence of human error
 - Not one test, but hundreds
- Testing numerical software challenging:
 - Generation of input data
 - Avoiding test failure due to numerics

Ideas and outlook

- Current implementation specific to `linalgwrap`
- Generator procedures could be generalised
- More systematic approaches useful:
 - Exploit mathematical theorems
 - Experiment with some algorithms

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References

- https://github.com/emil-e/rapidcheck/blob/master/doc/user_guide.md
- <https://en.wikipedia.org/wiki/QuickCheck>
- <https://linalgwrap.org>
- <https://linalgwrap.org/krimis>



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