An advanced C++ framework for numerical integration of ordinary differential equations

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Outline

- Programming issues in scientific computing
- 2 odeint introduction
- 3 odeint internals
- Outlook and conclusion

The interface problem in C/C++

- Many frameworks exist to do numerical computations.
- Data has to be stored in containers or collections.
- **GSL**: gsl_vector, gsl_matrix
- NR: pointers with Fortran-style indexing
- Blitz++, MTL4, boost::ublas
- QT: QVector, wxWidgets: wxArray, MFC: CArray

But: All books on C++ recommend the use of the STL containers std::vector, std::list,...

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Theoretical solution of the interface mess

GoF Design Pattern: Adaptor, also known as Wrapper

Gamma, Holm, Johnson, Vlissides: Design Patterns, Elements of Reuseable Object-Oriented Software, 1998.

Example

```
void CrankNicolsonEvolution::prepareVector(gsl vector complex* phi) {
 gsl vector complex* phi temp = gsl vector complex alloc(dim);
 //we need a copy of phi for this
 gsl vector complex memcpv(phi temp, phi):
 for( int i=1; i < dim-1; i++ ) {
   //phi n = phi n - i*dt/2 * (phi n-1 + phi n+1 + pot[n]*phi n)
   gsl vector complex set(phi, i, gsl complex add(
               gsl vector complex get(phi temp, i),
               gsl complex mul imag(
                  gsl complex add(
                     gsl complex add( gsl vector complex get(phi temp, i-1),
                                   gsl_vector_complex_get(phi_temp, i+1)),
                     gsl complex mul real( gsl vector complex get(phi temp, i),
                                           potential[i] )),
                  -dt/2.0))):
  if(periodic) {
   //periodic boundaries: i=0
   gsl vector complex_set(phi, 0, gsl_complex_add(
               gsl vector complex get(phi temp, 0),
               gsl complex mul imag(
                  gsl complex add(
                     gsl complex add( gsl vector complex get(phi temp, dim-1),
                                      gsl vector complex get(phi temp, 1)),
                     gsl complex mul real( gsl vector complex get(phi temp, θ),
                                           potential[0] )).
                  -dt/2.0))):
   //periodic boundaries: i=dim-1
   gsl vector complex set(phi, dim-1, gsl complex add(
               gsl vector complex get(phi temp, dim-1),
               gsl complex mul imag(
                  gsl complex add(
                     gsl complex add( gsl vector complex get(phi temp, dim-2),
                                      gsl vector complex get(phi temp, 0)),
                     gsl complex mul real(gsl vector complex get(phi temp, dim-1),
                                           potential[dim-1] )),
                  -dt/2.0))):
  } else {
```

Scalability of your algorithm

How to run your algorithm?

- Single machine, single CPU
- Single machine, multiple CPU's (OpenMP, threads, ...)
- Multiple machines (MPI)
- GPU (Cuda, Thrust)

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- Build-in data types double, complex<double>
- Arbitrary precision types GMP, MPFR
- Vectorial data types float2d, float3d

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Theoretical solution

GoF Design Pattern: Strategy, also known as Policy

Gamma, Holm, Johnson, Vlissides: Design Patterns, Elements of Reuseable Object-Oriented Software, 1998.

Numerical integration of ODEs

Find a numerical solution of an ODE an its initial value problem

$$\dot{x}=f(x,t)$$
, $x(t=0)=x_0$

Example: Explicit Euler

$$x(t + \Delta t) = x(t) + \Delta t \ f(x(t), t) + \mathcal{O}(\Delta t^2)$$

General scheme of order s

$$x(t)\mapsto x(t+\Delta t)$$
 , or $x(t+\Delta t)=\mathcal{F}_t x(t)+\mathcal{O}(\Delta t^{s+1})$

Solving ordinary differential equations in C++

Open source

Boost license – do whatever you want do to with it

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https://github.com/headmyshoulder/odeint-v2

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Modern C++

- Generic programming, functional programming
- Heavy use of the C++ template system
- Fast, easy-to-use and extendable.
- Container independent
- Scalable

First example – Lorenz system

```
#include <boost/numeric/odeint.hpp>
#include <tr1/array>
using namespace boost::numeric::odeint;
typedef std::tr1::array<double,3> state type;
void lorenz(const state type &x, state type &dxdt, double t) {
    // ...
int main(int argc, char **argv) {
    state_type x = \{\{10.0, 10.0, 10.0\}\};
    typedef dense output runge kutta<
        controlled runge kutta<
            runge_kutta_dopri5<state_type> > > stepper_type;
    integrate const(stepper_type(), lorenz, x, 0.0, 10.0, 0.01);
   return 0;
```

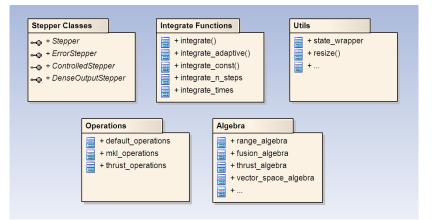
- The r.h.s. of the ODE is a simple function
- Methods with dense-output and/or step-size control
- Integrate functions
- Templates

Second example – Fermi-Pasta-Ulam lattice

```
typedef std::vector<double> state type:
struct fpu {
   double m beta:
    fpu(double beta) : m_beta(beta) { }
    void operator()(const state type &g, state type &dpdt) const {
        // ...
};
void statistics_observer ( const state_type &x , double t ) {
    // write the statistics
int main(int argc, char **argv) {
    state type q(256),p(256);
    // initialize q,p
    integrate_const(symplectic_rkn_sb3a_mclachlan<state_type>(), fpu(1.0),
         make pair(q,p), 0.0, 10.0, 0.01, statistics observer());
    return 0:
```

- Symplectic solver
- The ODE is now a functor, it can have parameters.
- Automatic memory managements
- Observer

Structure of odeint



User provides

$$y_i = f_i(x(t), t)$$

odeint provides

$$x_i(t + \Delta t) = x_i(t) + \Delta t \cdot y_i$$

(In general vector operations like $z_i = a_1 x_{1,i} + a_2 x_{2,i} + ...$)

Instantiation

```
euler<state_type, value_type, deriv_type,
    time_type, algebra, operations > stepper;
```

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Instantiation

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```

All elements for container independence and scalability are already included in this line!

$$y_i = f_i(x(t))$$

$$x_i(t + \Delta t) = x_i(t) + \Delta t \cdot y_i$$

```
euler<state_type, value_type, deriv_type,
    time_type, algebra, operations > stepper;
```

General goal: Separation

- of how an vector is iterated
- of how the basic computations are performed

from the stepper

$$y_i = f_i(x(t))$$

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```
euler<state_type, value_type, deriv_type,
   time_type, algebra, operations > stepper;
```

Data types

- state_type the type of x
- value_type the basic numeric type, e.g. double
- deriv_type the type of y
- time_type the type of t, Δt

$$y_i = f_i(x(t))$$

$$x_i(t + \Delta t) = x_i(t) + \Delta t \cdot y_i$$

```
euler<state_type, value_type, deriv_type,
    time_type, algebra, operations > stepper;
```

Algebra do the iteration

Algebra must be a class with public methods

- for_each1(x,op) Performs $op(x_i)$ for all i
- for_each2(x1,x2,op) Performs $op(x1_i,x2_i)$ for all i
- ...

$$y_i = f_i(x(t))$$

$$x_i(t + \Delta t) = x_i(t) + \Delta t \cdot y_i$$

```
euler<state_type, value_type, deriv_type,
    time_type, algebra, operations > stepper;
```

Operations do the basic computation

Operations must be a class with the public classes (functors)

- scale_sum1 Calculates $x = a1 \cdot y1$
- scale_sum2 Calculates $x = a1 \cdot y1 + a2 \cdot y2$
- ...

$$y_i = f_i(x(t))$$

$$x_i(t + \Delta t) = x_i(t) + \Delta t \cdot y_i$$

```
euler<state_type, value_type, deriv_type,
   time_type, algebra, operations > stepper;
```

All together

```
m_algebra.for_each3(xnew ,xold, y ,
operations_type::scale_sum2<value_type,time_type>(1.0,dt));
```

Stepper concepts

Concepts

"... In generic programming, a concept is a description of supported operations on a type..."

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odeint provides

Stepper concept

```
stepper.do_step(sys,x,t,dt);
```

ErrorStepper concept

```
stepper.do_step(sys,x,t,dt,xerr);
```

ControlledStepper concept

```
stepper.try_step(sys,x,t,dt);
```

DenseOutputStepper concept

```
stepper.do_step(sys);
stepper.calc_state(t,x);
```

Supported methods

Method Euler Runge-Kutta 4 Runge-Kutta Cash-Karp Runge-Kutta Fehlberg Runge-Kutta Dormand-Prince	Class name euler runge_kutta4 runge_kutta_cash_karp54 runge_kutta_runge_fehlberg78 runge_kutta_dopri5	Concept SD S SE SE SED
Runge-Kutta controller	controlled_runge_kutta	C
Runge-Kutta dense output	dense_output_runge_kutta	D
Symplectic Euler	symplectic_euler	S
Symplectic RKN	symplectic_rkn_sb3a_mclachlan	S
Rosenbrock 4	rosenbrock4	ECD
Implicit Euler	implicit_euler	S
Adams-Bashforth-Moulton Bulirsch-Stoer	adams_bashforth_moulton bulirsch_stoer	S CD

S – fulfills stepper concept

E – fulfills error stepper concept

C – fulfills controlled stepper concept

D – fulfills dense output stepper concept

Integrate functions

- integrate_const
- integrate_adaptive
- integrate_times
- integrate_n_steps

Perform many steps, use all features of the underlying method

Integrate functions

- integrate_const
- integrate_adaptive
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Perform many steps, use all features of the underlying method

An additional observer can be called

```
integrate_const(stepper, sys, x, t_start,
t_end, dt, obs);
```

More internals

- Header-only, no linking → powerful compiler optimization
- Memory allocation is managed internally
- No virtual inheritance, no virtual functions are called
- Different container types are supported, for example
 - STL containers (vector, list, map, tr1::array)
 - MTL4 matrix types, blitz++ arrays, Boost.Ublas matrix types
 - thrust::device_vector
 - Fancy types, like Boost.Units
 - ANY type you like
- Explicit Runge-Kutta-steppers are implemented with a new template-metaprogramming method
- Different operations and algebras are supported
 - MKL
 - Thrust
 - gsl

ODEs on GPUs

Graphical processing units (GPUs) are able to perform up to 10⁶ operations at once in parallel

Frameworks

- CUDA from NVIDIA
- OpenCL
- Thrust a STL-like library for CUDA and OpenMP

Applications:

- Parameter studies
- Large systems, like ensembles or one- or two dimensional lattices
- Discretizations of PDEs

odeint supports CUDA, through Thrust

Example: Parameter study of the Lorenz system

```
typedef thrust::device vector<double> state type;
typedef runge kutta4<state type ,value type ,state type ,value type ,
     thrust algebra .thrust operations > stepper type;
struct lorenz system {
    lorenz_system(size_t N ,const state_type &beta)
    : m N(N) , m beta(beta) {}
    void operator()( const state_type &x , state_type &dxdt , double t ){
        // ..
    size t m N;
    const state type &m beta;
};
int main ( int arc , char* argv[] )
    const size t N = 1024:
    vector<value type> beta host(N);
    for( size_t i=0 ; i<N ; ++i )
        beta host[i] = 56.0 + \text{value type}(i) * (56.0) / \text{value type}(N - 1)
              );
    state type beta = beta_host;
    state_type x(3 * N, 10.0);
    integrate const( stepper type() , lorenz(N,beta) , x , 0.0 , 10.0 , 0.01
         );
    return 0;
```

Conclusion

- odeint provides a fast, flexible and easy-to-use C++ library for numerical integration of ODEs.
- Its container independence is a large advantage over existing libraries.
- Scalable
- Generic programming is the main programming technique.

Outlook

- Submission to the boost libraries
- Dynamical system classes for easy implementation of interacting dynamical systems
- More methods: implicit methods and multistep methods.
- Implementation of the Taylor series method

```
taylor_fixed_order< 25 , 3 > taylor_type stepper;

stepper.do_step(
   fusion::make_vector
   (
       sigma * ( arg2 - arg1 ) ,
       R * arg1 - arg2 - arg1 * arg3 ,
       arg1 * arg2 - b * arg3
   ) , x , t , dt );
```

Resources

An article about the used techniques exists at

http://www.codeproject.com/KB/recipes/odeint.aspx

Download and documentation

http://headmyshoulder.github.com/odeint-v2/

Development

https://github.com/headmyshoulder/odeint-v2

Contributions and feedback

are highly welcome