Solving ordinary differential equations in C++

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Outline

- Introduction
- 2 Tutorial
- Technical details
- Discussion

Newtons equations

Newtons equations

Reaction and relaxation equations (i.e. blood alcohol content)

Newtons equations



Reaction and relaxation equations (i.e. blood alcohol content)

Granular systems



Newtons equations

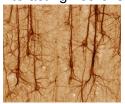


Reaction and relaxation equations (i.e. blood alcohol content)

Granular systems



Interacting neurons



Newtons equations

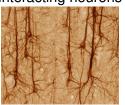


Reaction and relaxation equations (i.e. blood alcohol content)

Granular systems



Interacting neurons



- Many examples in physics, biology, chemistry, social sciences
- Fundamental in mathematical modelling

What is an ODE?

$$rac{\mathrm{d}x(t)}{\mathrm{d}t} = fig(x(t),tig)$$
 short form $\dot{x} = f(x,t)$

- x(t) − dependent variable
- *t* indenpendent variable (time)
- f(x, t) defines the ODE

Initial Value Problem (IVP):

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

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++

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Modern C++

- Generic programming, functional programming
- Fast, easy-to-use and extendable.
- Container independent
- Portable

Who uses odeint

NetEvo



OMPL – Open Motion Planning Library

Motivation: 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

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GoF Design Pattern: Adaptor, also known as Wrapper

Alternative

Generic, container independent algorithms

Portability 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, OpenCL)

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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

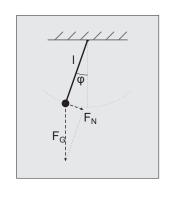
GoF Design Pattern: Strategy, also known as Policy

Alternative

Generic algorithms

Lets step into odeint

- Introduction
- 2 Tutorial
- Technical details
- 4 Discussion



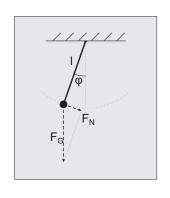
Newtons law: ma = F

Acceleration: $a = I\ddot{\varphi}$

Force: $F = F_N = -mg \sin \varphi$

$$\Longrightarrow$$
 ODE for φ

$$\ddot{\varphi} = -g/I\sin\varphi = -\omega_0^2\sin\varphi$$



$$\ddot{\varphi} = -\omega_0^2 \sin \varphi$$

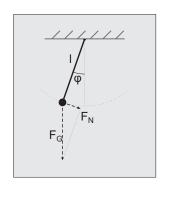
Small angle: $\sin \varphi \approx \varphi$

Harmonic oscillator $\ddot{\varphi} = -\omega_0^2 \varphi$

Analytic solution:

$$\varphi = A\cos\omega_0 t + B\sin\omega_0 t$$

Determine A and B from initial condition



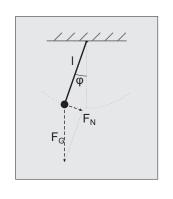
Full equation: $\ddot{\varphi} = -\omega_0^2 \sin \varphi$

Pendulum with friction and external driving:

$$\ddot{\varphi} = -\omega_0^2 \sin \varphi - \mu \dot{\varphi} + \varepsilon \sin \omega_E t$$

No analytic solution is known

 \Longrightarrow Solve this equation numerically.



$$\ddot{\varphi} = -\omega_0^2 \sin \varphi - \mu \dot{\varphi} + \varepsilon \sin \omega_E t$$

Create a first order ODE

$$x_1 = \varphi$$
, $x_2 = \dot{\varphi}$

$$\dot{X_1} = X_2$$

$$\dot{x_2} = -\omega_0 \sin x_1 - \mu x_2 + \varepsilon \sin \omega_E t$$

 x_1 and x_2 are the state space variables

```
#include <boost/numeric/odeint.hpp>
namespace odeint = boost::numeric::odeint;
```

$$\dot{x_1} = x_2$$
, $\dot{x_2} = -\omega_0 \sin x_1 - \mu x_2 + \varepsilon \sin \omega_E t$

typedef std::array<double,2> state_type;

```
\dot{x_1} = x_2, \, \dot{x_2} = -\omega_0^2 \sin x_1 - \mu x_2 + \varepsilon \sin \omega_E t
```

```
struct pendulum
 double m_mu, m_omega, m_eps;
 pendulum (double mu, double omega, double eps)
  : m mu(mu), m omega(omega), m eps(eps) { }
 void operator()(const state_type &x,
     state type &dxdt, double t) const
    dxdt[0] = x[1];
    dxdt[1] = -\sin(x[0]) - m mu * x[1] +
        m eps * sin(m omega*t);
```

$$\varphi(0)=1$$
 , $\dot{\varphi}(0)=0$

```
odeint::rk4< state_type > rk4;
pendulum p( 0.1 , 1.05 , 1.5 );

state_type x = {{ 1.0 , 0.0 }};
double t = 0.0;

const double dt = 0.01;
rk4.do_step( p , x , t , dt );
t += dt;
```

$$x(0) \mapsto x(\Delta t)$$

```
std::cout<<t<" "<< x[0]<<" "<<x[1]<<"\n";
for( size_t i=0 ; i<10 ; ++i )
{
   rk4.do_step( p , x , t , dt );
   t += dt;
   std::cout<<t<<" "<< x[0]<<" "<<x[1]<<"\n";
}</pre>
```

$$x(0)\mapsto x(\Delta t)\mapsto x(2\Delta t)\mapsto x(3\Delta)\mapsto\dots$$
Grafik einfuegen

Simulation

Oscillator:
$$\mu = 0$$
 , $\omega_E = 0$, $\varepsilon = 0$

Damped oscillator:
$$\mu = 0.1$$
 , $\omega_E = 0$, $\varepsilon = 0$

Damped, driven oscillator: $\mu = 0.1$, $\omega_E = 1.05$, $\varepsilon = 1.5$

Different Steppers

```
runge_kutta_fehlberg78< state_type > s;
```

```
runge_kutta_dopri5< state_type > s;
```

Symplectic steppers (for Hamiltonian systems)

```
symplectic_rkn_sb3a_mclachlan< state_type > s;
```

Implicit steppers (for stiff systems)

```
rosenbrock4< double > s;
```

These steppers perform one step with constant step size!

Controlled steppers – Step size control

insert graphic

Controlled steppers

```
auto s = make_controlled(1.0e-6,1.0e6,
  runge_kutta_fehlberg78<state_type>() );
controlled_step_result r =
  s.try_step(ode,x,t,dt);
```

Tries to perform the step and updates x, t, and dt!

It works because Runge-Kutta-Fehlberg has error estimation:

```
runge_kutta_fehlberg78<state_type> s;
s.do_step(ode,x,t,dt,xerr);
```

Controlled steppers

```
auto s = make controlled(1.0e-6, 1.0e6,
  runge_kutta_fehlberg78<state_type>() );
while ( t < t end )
  controlled_step_result res
    = s.try_step(ode,x,t,dt);
  while( res != success )
    res = s.try_step(ode, x, t, dt);
```

Non-trivial time-stepping logic

Use integrate functions!

Observer: Callable object obs(x,t)

Example (using Boost.Phoenix):

```
integrate_adaptive(s,ode,x,t_start,t_end,dt,
    cout<< arg1[0] << " " << arg1[1] << "\n" );</pre>
```

More integrate versions:

```
integrate_const, integrate_times,...
```

integrate_const(s,ode,x,t,dt,obs);

Grafik with problem and solution

Dense output

```
auto s = make_dense_output( 1.0e-6 , 1.0e-6 ,
    runge_kutta_dopri5< state_type >() );
integrate_const( s , p , x , t , dt );
```

Interpolation between two steps with same precision as the original stepper!

Grafik!

More steppers

Stepper Concepts: Stepper, ErrorStepper, ControlledStepper, DenseOutputStepper

Stepper types:

- Implicit implicit_euler, rosenbrock4
- Symplectic symplectic_rkn_sb3a_mclachlan
- Predictor-Corrector adams_bashforth_moulton
- Extrapolation bulirsch_stoer
- Multistep methods adams_bashforth_moulton

Some of them have step-size control and dense-output!

Small summary

- Very easy example harmonic oscillator
- Basic features of odeint
- Different stepper Controlled steppers, Dense output steppers
- Integrate functions

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- Very easy example harmonic oscillator
- Basic features of odeint
- Different stepper Controlled steppers, Dense output steppers
- Integrate functions

Now, lets look at the advanced features!

Lattice systems

Lattice systems

Discretiztations of PDEs

Lattice systems

Discretiztations of PDEs

Granular systems



Lattice systems

Granular systems



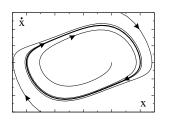
Discretiztations of PDEs

ODEs on graphs



High-Performance-Computing

Phase oscillator lattices



Any oscillator can be described by one variable, its phase.

Trivial dynamics: $\dot{\varphi} = \omega \varphi$ Coupled phase oscillators Neurosciences

Heart dynamics

Synchronization

Any weakly perturbed oscillator system

$$\dot{\varphi}_{k} = \omega_{k} \varphi_{k} + q(\varphi_{k+1}, \varphi_{k}) + q(\varphi_{k}, \varphi_{k-1})$$

Phase compacton lattice

$$\dot{\varphi}_k = \cos\varphi_{k+1} - \cos\varphi_{k-1}$$

State space contains *N* variables

```
typedef std::vector<double> state_type;
```

Animation

Space-time plot for visualization of compactons and chaos

Ensemble of phase oscillators

$$\dot{\varphi}_{k} = \omega_{k} + \sum_{I} \sin(\varphi_{I} - \varphi_{k})$$

Synchronization – all oscillator oscillates with the same frequency

Synchronized state $\varphi_k = \omega_S t + \varphi_{0,k}$

Ensemble of phase oscillators

```
typedef std::vector<double> state_type;
struct ensemble
    state_type m_omega, m_eps;
    ensemble (size_t n, double eps)
    : m_omega(n,0.0),m_eps(eps)
        create frequencies();
    void create frequencies() { ... }
    void operator() (const state type &x,
       state type &dxdt, double t) const
```

Solving ODEs with CUDA using Thrust

Thrust is a parallel algorithms library which resembles the C++ Standard Template Library (STL). Thrust's high-level interface greatly enhances developer productivity while enabling performance portability between GPUs and multicore CPUs. Interoperability with established technologies (such as CUDA, TBB and OpenMP) facilitates integration with existing software. Develop high-performance applications rapidly with Thrust!



Solving ODEs with CUDA using thrust

Applications and use cases for GPUs:

- Large systems, discretizations of PDEs, lattice systems, granular systems, etc.
- Parameter studies, solve many ODEs in parallel with different parameters
- Initial value studies, solve the same ODE with many different initial conditions in parallel

Lorenz system - Parameter study

$$\dot{x} = \sigma(y - x)$$
 $\dot{y} = Rx - y - xz$ $\dot{z} = -bz + xy$
Standard parameters $\sigma = 10$, $R = 28$, $b = 8/3$

Deterministic chaos – Butterfly effect Picture of Lorenz system

Lorenz system – Parameter study

$$\dot{x} = \sigma(y - x)$$
 $\dot{y} = Rx - y - xz$ $\dot{z} = -bz + xy$

Does one observ chaos over the whole parameter range?

Lyapunov exponents:

- Measure of chaos
- Perturbations of the original system

Vary *R* from 0 to 50 and calculate the Lyapunov exponents!

Use CUDA and Thrust!

odeint's algebras and operations

Euler method

$$x_i(t + \Delta t) = x_i(t) + \Delta t * f_i(x)$$

Algebras perform the iteration over *i* and operation the elementary addition.

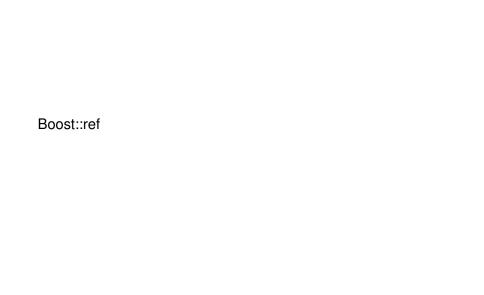
Algebras and operations enter the stepper as template parameters

```
typedef runge_kutta4<state_type, value_type,
  deriv_type, time_type,
  algebra, operations, resize_policy> stepper;
```

- default_operations
- range_algebra Boost.Ranges
- vector_space_algebra Passes the state directly to the operations
- fusion_algebra For compile time sequences, like std :: tuple < double, double >
- thrust_algebra and thrust_algebra for thrust

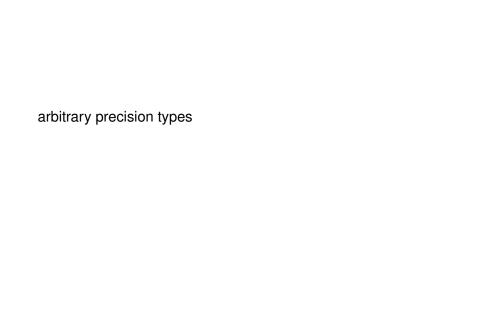
Thrust example for Lorenz system,
Implementation of the system function

More advanced features, die themen können auch auf mehreren folien zusammengefasst werden

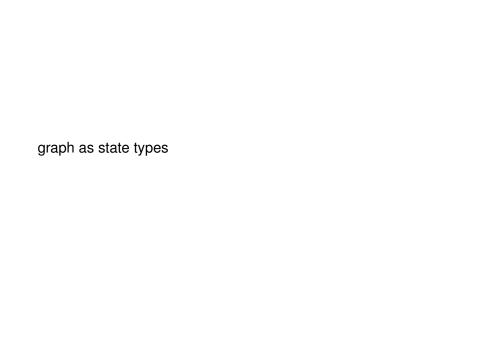


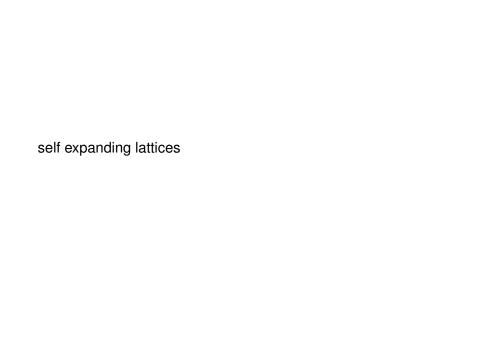
boost::range

complex state types, vielleicht auch nicht









Introduction

2 Tutorial

Technical details

Discussion

Independent Algorithms

Goal

Container- and computation-independent implementation of the numerical algorithms.

Benefit

High flexibility and applicability, ODEINT can be used for virtually any formulation of an ODE.

Approach

Detatch the algorithm from memory management and computation detail and make each part interchangeable.

Mathematical Algorithm

Typical mathematical computation to calculate the solution of an ODE $(\vec{x} = \vec{f}(\vec{x}, t))$:

$$\vec{F}_{1} = \vec{f}(\vec{x}_{0}, t_{0})
\vec{x}' = \vec{x}_{0} + a_{21} \cdot \Delta t \cdot \vec{F}_{1}
\vec{F}_{2} = \vec{f}(\vec{x}', t_{0} + c_{1} \cdot \Delta t)
\vec{x}' = \vec{x}_{0} + a_{31} \cdot \Delta t \cdot \vec{F}_{1} + a_{32} \cdot \Delta t \cdot \vec{F}_{2}
\vdots
\vec{x}_{1} = \vec{x}_{0} + b_{1} \cdot \Delta t \cdot \vec{F}_{1} + \dots + b_{n} \cdot \Delta t \cdot \vec{F}_{n}$$

Strucutural Requirements

$$\vec{F}_1 = \vec{f}(\vec{x}_0, t_0)$$
 $\vec{x}' = \vec{x}_0 + a_{21} \cdot \Delta t \cdot \vec{F}_1$

Types:

- vector type, mostly, but not neccessarily, some container like vector<double> (actually we have state_type and deriv_type)
- time type, usually double, but might be a multi-precision type
- value type, most likely the same as time type

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- value type, most likely the same as time type

Function Call:

```
void rhs( const vector_type &x , vector_type &
    dxdt , const time_type t )
{ /* user defined */ }
rhs( x0 , F1 , t ); //memory allocation for F1?
```

Memory allocation for temporary results (F, x')

Computational Requirements

$$\vec{x}_1 = \vec{x}_0 + b_1 \cdot \Delta t \cdot \vec{F}_1 + \dots + b_s \cdot \Delta t \cdot \vec{F}_s$$

- vector-vector addition
- scalar-scalar multiplication
- scalar-vector multiplication

 $(\longrightarrow \text{vector space})$

Type Declarations

Tell ODEINT which types your are working with:

Reasonable standard values for the template parameters allows for:

```
typedef runge_kutta4<state_type> stepper_type;
```

Memory Allocation / Resizing

Two possible situations: dynamic size / fixed size vector_type

dynamic size - memory allocation required

- e.g. vector<double>
- declare type as resizeable
- specialize resize template
- USe initially_resizer
 or always_resizer in
 stepper algorithm

fixed size - memory allocation not required

- e.g. array<double, N>
- declare type as not resizeable
- that's it

Declare Resizeability

```
/* by default any type is not resizable */
template< class Container >
struct is resizeable
   typedef boost::false type type;
   const static bool value = type::value;
};
/* specialization for std::vector */
template < class T, class A >
struct is_resizeable< std::vector< T , A >>
   typedef boost::true_type type;
   const static bool value = type::value;
};
```

To use a new dynamic sized type, this has to be specialized by the user.

Tell ODEINT how to resize

Again: only required if

```
is_resizeable<state_type>::type == boost::true_type.
```

Class Template responsible for resizing:

```
template< class StateOut , class StateIn >
struct resize_impl
{
    /* standard implementation */
    static void resize( StateOut &x1 , const
        StateIn &x2 )
    {
        x1.resize( boost::size( x2 ) );
    }
};
```

For anything that does not support boost::size or resize the user must provide a specialization.

Scalar Computations

For the scalar types we require the following: Assume:

```
time_type t , dt;
value_type a1 , a2 , c;
```

Valid Expressions:

- a1 = static_cast< value_type > (1)
- a1*a2
- a1/a2
- t + c*dt
- t + dt/c
- t += dt

Remember: $\vec{x}_1 = \vec{x}_0 + b_1 \cdot \Delta t \cdot \vec{F}_1 + \cdots + b_s \cdot \Delta t \cdot \vec{F}_s$

Split into two parts:

- 1. Algebra: responsible for iteration over vector elements
- 2. Operations: does the mathematical computation on the elements

Very similar to std::for_each, ODEINT uses something like:

Remember: $\vec{x}_1 = \vec{x}_0 + b_1 \cdot \Delta t \cdot \vec{F}_1 + \cdots + b_s \cdot \Delta t \cdot \vec{F}_s$ Split into two parts:

- 1. Algebra: responsible for iteration over vector elements
- 2. Operations: does the mathematical computation on the elements

Very similar to std::for_each, ODEINT uses something like:

```
Algebra algebra;

algebra.for_each3( x1 , x0 , F1 ,
Operations::scale_sum3( 1.0, b1*dt );
```

The types Algebra and Operations are template parameters of the steppers, hence exchangeable.

```
time_type dt;
state_type x1, x2, ...
algebra_type algebra;
```

Algebra has to have defined the following member functions:

```
algebra.for_each1( x1 , unary_operation );
algebra.for_each2( x1, x2, binary_operation );
algebra.for_each3( ... );
:
algebra.for_each15( ... , fifteen_ary_op );
```

```
time_type dt;
state_type x1, x2, ...
algebra_type algebra;
```

Algebra has to have defined the following member functions:

```
algebra.for_each1( x1 , unary_operation );
algebra.for_each2( x1, x2, binary_operation );
algebra.for_each3( ... );
:
algebra.for_each15( ... , fifteen_ary_op );
```

ODEINT takes the operations from the class Operations.

Operations

Operations is a class with the following member classes:

- scale
- scale_sum1
- scale_sum2

÷

• scale_sum14

These classes need a constructor and ()-operator that works together with the algebra:

```
value_type b1, b2;
time_type dt;
algebra.for_each4( x1 , x0 , F1 ,
    Operations::scale_sum3( 1.0, b1*dt );
```

This computes: $\vec{x}_1 = 1.0 \cdot \vec{x}_0 + b_1 \Delta t \cdot \vec{F}_1$.

Example Implementation: range_algebra

```
struct range algebra {
template < class S1 , class S2 , class S3 , class Op >
 static void for each2(S1 &s1, S2 &s2, S3 &s3, Op op)
   detail::for_each2( boost::begin(s1), boost::end(s1),
                      boost::begin(s2), boost::begin(s3),
                      op );
};
namespace detail {
template < class Iter1, class Iter2, Iter3, class Op >
void for each2( Iter1 first1, Iter1 last1,
                 Iter2 first2, Iter3 first3, Op op )
     for( ; first1 != last1 ; )
         op(*first1++,*first2++,*first3++);
```

Example Implementation: default_operations

```
template < class Fac1 , class Fac2 >
struct scale sum2
  const Fac1 m alpha1;
  const Fac2 m alpha2;
  scale_sum2( Fac1 &alpha1 , Fac2 &alpha2 )
    : m alpha1 ( alpha1 ) , m alpha2 ( alpha2 )
 template< class T1 , class T2 , class T3 >
 void operator() ( T1 &t1 , const T2 &t2 ,
    const T3 &t3 )
  \{ t1 = m_alpha1 * t2 + m_alpha2 * t3; \}
 typedef void result_type;
};
```

range_algebra & default_operations can be used with any Container that supports Boost.Range and whose container::value_type supports operators +, *.

For example vector< double >:

As these are also the default values, this can be shortened:

```
typedef runge_kutta4<state_type> stepper_type;
```

range_algebra & default_operations work also with

- vector< complex<double> >
- list< double >
- array< double , N >

range_algebra & default_operations work also with

- vector< complex<double> >
- list< double >
- array< double , N >

What about

- Ublas vector
- trivial state types like double
- generally: state_type that support operators +, *

→ vector_space_algebra!

vector_space_algebra

```
struct vector_space_algebra {
template< class S1 , class S2 , class S3 ,
    class Op >
static void for_each3( S1 &s1 , S2 &s2 ,
                       S3 &s3 , Op op )
  op(s1,s2,s3);
```

- delegates operations directly to the state_type
- no iteration
- works together with default_operations with any state_type that supports operators +, *

Other Examples

```
fusion algebra: works with fusion::vector as state_type,
            (e.g. Boost.Units)
```

thrust algebra & thrust operations: Use thrust library to perform computation on CUDA graphic cards

nested algebra: can handle nested containers that support Boost.Range, e.g. vector < vector < double > >

See tutorial and documentation on www.odeint.com for more.

Other Examples

thrust_algebra & thrust_operations: Use thrust library to perform computation on CUDA graphic cards

nested_algebra: can handle nested containers that support
Boost.Range, e.g. vector< vector<double> >

See tutorial and documentation on www.odeint.com for more.

Important

Division into Algebra and Operations gives us great flexibility. However, State_Type, Algebra and Operations must coorporate to make ODEINT work!

Introduction

2 Tutorial

Technical details

Discussion