### 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

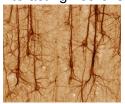


Reaction and relaxation equations (i.e. blood alcohol content)

Granular systems



Interacting neurons



Newtons equations

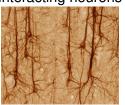


Reaction and relaxation equations (i.e. blood alcohol content)

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++

#### Open source

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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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#### 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

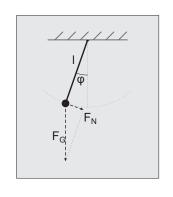
GoF Design Pattern: Strategy, also known as Policy

#### Alternative

Generic algorithms

# Lets step into odeint

- Introduction
- 2 Tutorial
- Technical details
- 4 Discussion



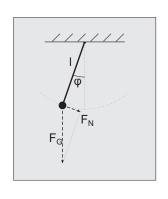
Newtons law: ma = F

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

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

$$\Longrightarrow$$
 ODE for  $\varphi$ 

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



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

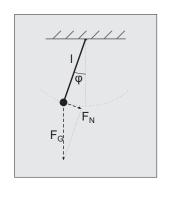
Small angle:  $\sin \varphi \approx \varphi$ 

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



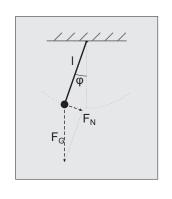
Full equation:  $\ddot{\varphi} = -\omega_0^2 \sin \varphi$ 

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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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



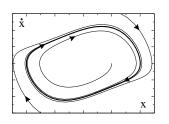
Discretiztations of PDEs

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.

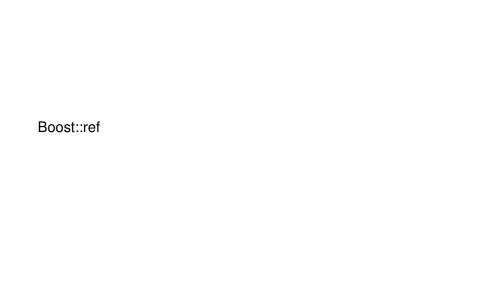
Algebras and operations enter the stepper as template parameters

```
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

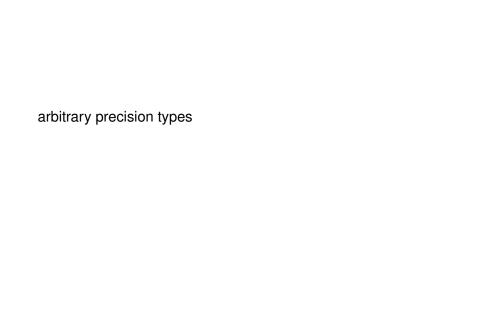
Thrust example for Lorenz system,
Implementation of the system function

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

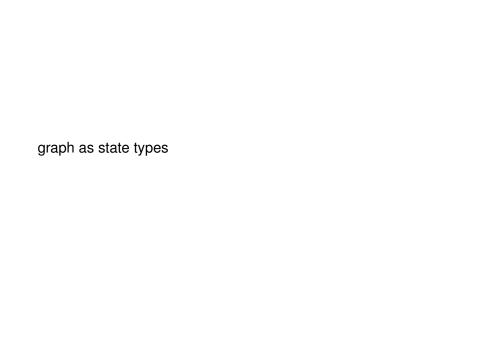


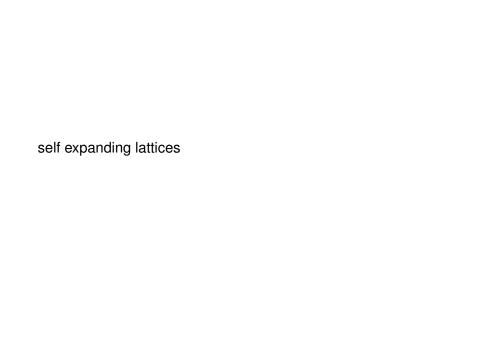
boost::range

complex state types, vielleicht auch nicht









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## 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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#### **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

(→ 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

# Vector Computations

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