

odeint

Solving ordinary differential equations in C++

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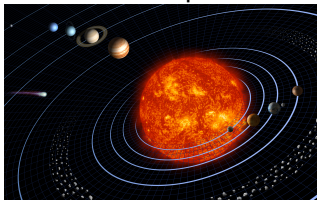


Outline

- 1 Introduction
- 2 Tutorial
- 3 Technical details
 - Independent Algorithms
 - Memory Management
 - Computation Backend
 - Benefits
- 4 Discussion

What is an ODE? – Examples

Newtons equations

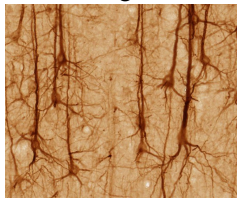


Reaction and relaxation equations (i.e. blood alcohol content)

Granular systems



Interacting neurons



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

What is an ODE?

$$\frac{dx(t)}{dt} = f(x(t), t) \quad \text{short form} \quad \dot{x} = f(x, t)$$

- $x(t)$ – dependent variable
- t – independent variable (time)
- $f(x, t)$ – defines the ODE

Initial Value Problem (IVP):

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

Find $x(t)$

Numerical integration of ODEs

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

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

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

$$x(t) \mapsto x(t + \Delta t) \quad , \text{ or}$$

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

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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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Which data types are used by your algorithm?

- 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

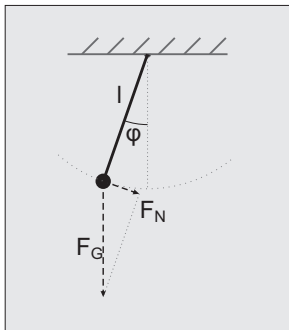
Alternative

Generic algorithms

Lets step into odeint

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Example – Pendulum



Pendulum – Newtons law

$$ma = F$$

Acceleration

$$a = l\ddot{\varphi}$$

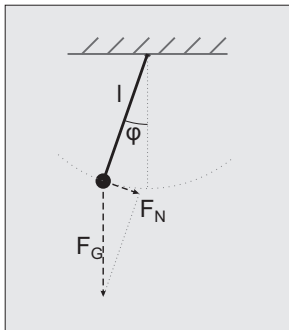
Force

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

Result in an ode for the angle

$$\ddot{\varphi} = -g/l \sin \varphi$$

Example – Pendulum



$$\ddot{\varphi} = -g/l \sin \varphi$$

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

Harmonic oscillator

$$\ddot{\varphi} = -g/l \varphi$$

An analytic solution is known

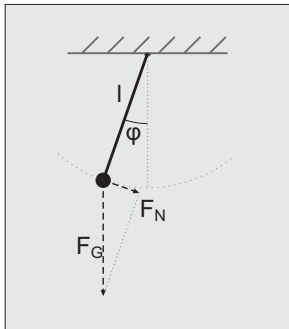
$$\varphi = A \cos \omega t + B \sin \omega t$$

Amplitude A and B must be determined from initial conditions:

$$\varphi(t=0) = \varphi_0, \dot{\varphi}(t=0) = \dot{\varphi}_0$$

$$B = \varphi_0, A = \dot{\varphi}_0 / \omega$$

Example – Pendulum

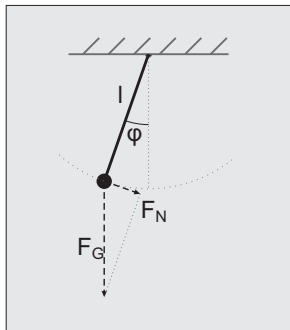


Full equation $\ddot{\varphi} = g/l \sin \varphi$
has also analytic solution Jacobi
elliptic function

Lets enhance the ODE, add fric-
tion and external driving

$\ddot{\varphi} = g/l \sin \varphi - \mu \dot{\varphi} + \varepsilon \sin \omega t$
No analytic solution is known.
We need to solve this equation
numerically.

Example – Pendulum



$$\ddot{\varphi} = g/l \sin \varphi - \mu \dot{\varphi} + \varepsilon \sin \omega t$$

Create a first order ODE

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

$$\dot{x}_1 = x_2, \dot{x}_2 = -g/l \sin x_1 - \mu x_2 + \varepsilon \sin \omega t$$

x_1 and x_2 are the state space variables.

Let's solve the pendulum example numerically

```
#include <boost/numeric/odeint.hpp>

namespace odeint = boost::numeric::odeint;
```

$$\dot{x}_1 = x_2, \dot{x}_2 = -g/l \sin x_1 - \mu x_2 + \varepsilon \sin \omega t$$

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

Let's solve the pendulum example numerically

$$\dot{x}_1 = x_2, \dot{x}_2 = -g/l \sin x_1 - \mu x_2 + \varepsilon \sin \omega t$$

```
struct pendulum
{
    double m_mu , m_omega , m_epsilon;

    pendulum( double mu , double omega , double epsilon )
    : m_mu( mu ) , m_omega( omega ) , m_epsilon( epsilon ) {
    }

    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_epsilon * sin
            ( m_omega * t );
    }
};
```

Let's solve the pendulum example numerically

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

```
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";
}
```

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

Simulation

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

$$\mu = 0.1, \omega_E = 0, \varepsilon = 0$$

$$\mu = 0.1, \omega_E = 1.05, \varepsilon = 1.5$$

Steppers

```
odeint::runge_kutta_fehlberg78< state_type > stepper;
```

```
odeint::runge_kutta_dopri5< state_type > stepper;
```

but controlled steppers are much better

Controlled steppers

insert graphic

```
auto stepper = make_controlled( 1.0e-6 , 1.0e6 , odeint::  
    runge_kutta_fehlberg78< state_type >() );  
odeint::controlled_step_result res = stepper.try_step( p ,  
    x , t , dt );
```

tries to perform the step and updates x , t , and dt

it works because runge kutta fehlberg has error estimation

Controlled steppers

```
auto stepper = make_controlled( 1.0e-6 , 1.0e6 , odeint::
    runge_kutta_fehlberg78< state_type >() );
while( t < t_end )
{
    odeint::controlled_step_result res = stepper.try_step( p
        , x , t , dt );
    while( res != odeint::success )
    {
        res = stepper.try_step( p , x , t , dt );
    }
}
```

Use integrate functions

```
integrate_adaptive( stepper , x , p , t_start , t_end , dt
    );
integrate_adaptive( stepper , x , p , t_start , t_end , dt
    , observer );
```

```
integrate_adaptive( stepper , p , x , t_start , t_end , dt
    ,
    std::cout << arg2 << "\t" << arg1[0] << "\t" << arg1[1]
        << "\n" );
```

`integrate_const`, `integrate_times`, ...

```
integrate_const( stepper , p , x , t , dt , observer );
```

problem with controlled stepper

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

More steppers

implicit, symplectic, predictor-corrector, multistep-methods

maybe small table

small summary (kann vielleicht auch wieder weg)

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

Now, advanced features

extended systems

- Lattice systems

High-Performance-Computing
zu jedem Punkt ein Bildchen

extended systems

- Lattice systems
- Discretizations of PDEs

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extended systems

- Lattice systems
- Discretizations of PDEs
- ODEs on Graphs
- granular systems

High-Performance-Computing
zu jedem Punkt ein Bildchen

Phase oscillator lattices

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

(Bild aus phd Talk)

Trivial dynamics: $\dot{\varphi} = \omega$

Vielleicht zusammenfuehren mit der naechsten Folie

Phase oscillator lattices

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 lattices

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

state space contains N variables

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

Animation with compactons and chaos

space-time plot for visualization of compactons and chaos

Ensemble of phase oscillators

$$\dot{\varphi}_k = \omega_k + \sum_l \sin(\varphi_l - \varphi_k)$$

Synchronization, all oscillator oscillates with the same frequency

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

Classical implementation

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

struct phase_ensemble
{
    state_type m_omega;
    double m_epsilon;

    phase_ensemble(size_t n, double g=1.0, double epsilon
                    =1.0)
        : m_omega(n, 0.0), m_epsilon(epsilon)
    {
        create_frequencies(g);
    }

    void create_frequencies(double g) { ... }

    void operator()(const state_type &x, state_type &dxdt,
                    double t) const
    {
        ...
    }
};
```

The ODE has now many parameters, use `boost::ref`
Vielleicht koennen diese beiden Folien weg

Solving ODEs with CUDA using thrust

What is 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

Where to use it

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

Lorenz system – Parameter study

$$\dot{x} = \sigma(y - x) \quad \dot{y} = Rx - y - xz \quad \dot{z} = -bz + xy \quad (1)$$

Standard parameters $\sigma = 10$, $R = 28$, $b = 8/3$ deterministic
chaos, butterfly effect
picture of Lorenz system

Lorenz system – Parameter study

Vary R from 0 to 50, for which parameters the system is chaotic?

Lyapunov exponents, perturbations of the original system

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::ref

boost::range

complex state types, vielleicht auch nicht

arbitrary precision types

matrices as state types

graph as state types

self expanding lattices

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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

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

Required Computations

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

$$\vec{F}_1 = \vec{f}(\vec{x}_0, t_0)$$

$$\vec{x}' = \vec{x}_0 + a_{21} \cdot \Delta t \cdot \vec{F}_1$$

$$\vec{F}_2 = \vec{f}(\vec{x}', t_0 + c_1 \cdot \Delta t)$$

$$\vec{x}' = \vec{x}_0 + a_{31} \cdot \Delta t \cdot \vec{F}_1 + a_{32} \cdot \Delta t \cdot \vec{F}_2$$

$$\vdots$$

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

Structural 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 necessarily, some container like `vector<double>`
- **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?
```

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