## Solving ordinary differential equations in C++

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October 12, 2012







## Outline

- Introduction
- 2 Tutorial
- Technical details
- 4 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

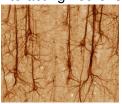


Reaction and relaxation equations (i.e. blood alcohol content)

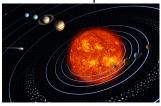
Granular systems



#### 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?

$$\frac{\mathrm{d}x(t)}{\mathrm{d}t} = f(x(t),t)$$
 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

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

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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#### 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

#### We want to solve ODEs

- using double, std::vector, std::array,...as state types.
- with complex numbers.
- on one, two and/or three-dimensional lattices.
- on Graphs.
- on graphic cards.
- with arbitrary precision types.

Container independent and portable algorithms are needed!

R

# Motivation: The interface problem in C/C++, SKIP!!

- 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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#### Theoretical solution of the interface mess

GoF Design Pattern: Adaptor, also known as Wrapper

#### Alternative

Generic, container independent algorithms

# Portability of your algorithm, SKIP!!

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

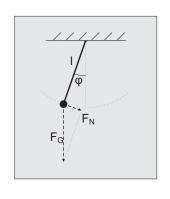
GoF Design Pattern: Strategy, also known as Policy

#### Alternative

Generic algorithms

# Lets step into odeint

- Introduction
- 2 Tutorial
- Technical details
- 4 Discussion



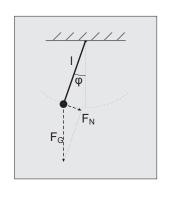
Newtons law: ma = F

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

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

 $\Longrightarrow$  ODE for  $\varphi$ 

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



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

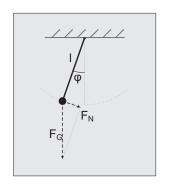
Small angle:  $\sin \varphi \approx \varphi$ 

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



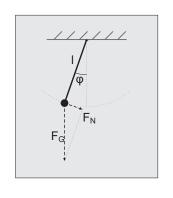
Full equation:  $\ddot{\varphi} = -\omega_0^2 \sin \varphi$ 

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);
```

$$\varphi(0) = x_1(0) = 1$$
,  $\dot{\varphi}(0) = x_2(0) = 0$ 

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

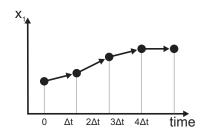
state_type x = {{ 1.0 , 0.0 }};
double t = 0.0;

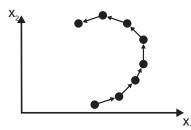
const double dt = 0.01;
rk4.do_step(p , x , t , dt );
t += dt;
```

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

```
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>
```

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





## Simulation

Oscillator: 
$$\mu = \mathbf{0}$$
 ,  $\omega_F = \mathbf{0}$  ,  $\varepsilon = \mathbf{0}$ 

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)

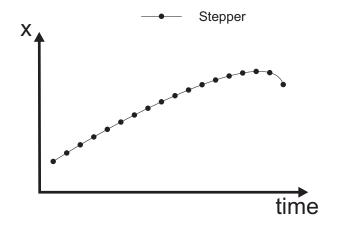
```
symplectic_rkn_sb3a_mclachlan< state_type > s;
```

### Implicit steppers (for stiff systems)

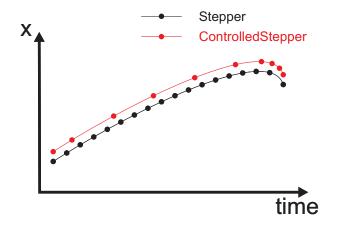
```
rosenbrock4< double > s;
```

### These steppers perform one step with constant step size!

# Controlled steppers – Step size control



# Controlled steppers – Step size control



# 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 )</pre>
  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,...
```

### Adaptive step size vs. constant step size

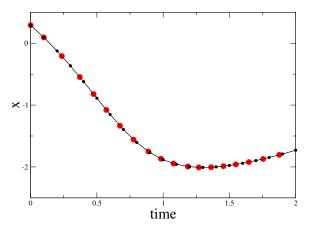
```
integrate_const(s,ode,x,t,dt,obs);
```

Performance loss when constant step size integration is needed

# Adaptive step size vs. constant step size

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

Performance loss when constant step size integration is needed

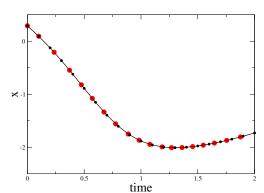


# Dense output stepper

#### Problem: Adaptive step size vs. constant step size integration

```
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 the same precision as the stepper!



#### 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

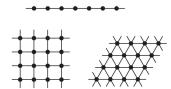
- Very easy example harmonic oscillator
- Basic features of odeint
- Different steppers Steppers, Error steppers, Controlled steppers, Dense output steppers
- Integrate functions

#### Small summary

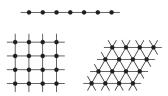
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Now, lets look at some advanced features!

#### Lattice systems



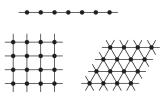
Lattice systems



Discretiztations of PDEs



Lattice systems

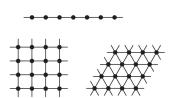




Discretiztations of PDEs



#### Lattice systems



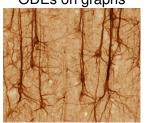
Granular systems



Discretiztations of PDEs

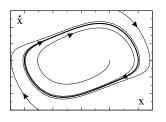


ODEs on graphs



**High-Performance-Computing** 

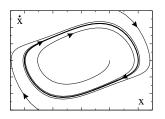
# Coupled phase oscillators



Any oscillator can be described by one variable – its phase!

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

### Coupled phase oscillators



Any oscillator can be described by one variable – its phase!

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

Interesting behaviour occurs if oscillators are coupled.

 Synchronization, oscillation death, phase chaos, pattern formation, . . .

Applications: Neurosciences, Heart dynamics, social systems

Any weakly coupled oscillator system

$$\dot{\varphi}_k = \omega_k + q(\varphi_{k+1}, \varphi_k) + q(\varphi_k, \varphi_{k-1})$$

### Phase compacton lattice

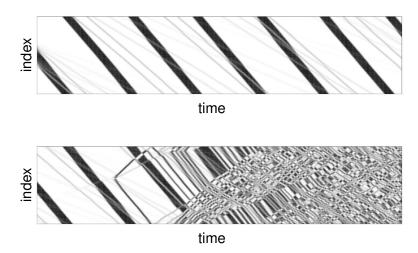
$$\dot{\varphi}_k = \cos\varphi_{k+1} - \cos\varphi_{k-1}$$

#### State space contains *N* variables

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

#### Animation

# Phase compacton lattice – Transition to chaos



### Ensemble of phase oscillators - SKIP

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

**Synchronization** – all oscillator oscillates with the same frequency

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

# Ensemble of phase oscillators – SKIP

```
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

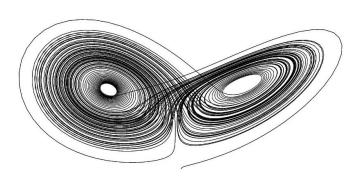
#### Applications and use cases for GPUs:

- 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 – Deterministic chaos

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

Perturbations grow exponentially fast – Butterfly effect



### 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!** 

# Intermezzo: Algebras and operations

#### Euler method

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

- Algebras perform the iteration over i
- Operations perform the elementary addition.

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

### Intermezzo: Algebras and operations

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

- default\_operations
- range\_algebra Boost.Ranges
- vector\_space\_algebra Passes the state directly to the operations
- fusion\_algebra Compile-time sequences, like std::tuple< double , double >
- thrust\_algebra and thrust\_algebra Thrust

### Calculate an ensemble of Lorenz systems

```
typedef thrust::device_vector<double> state_type;
typedef runge_kutta4<state_type, double, state_type, double,
    thrust_algebra, thrust_operations, resizer> stepper;

state_type x( N );
// initialize x
integrate_const( stepper() , lorenz_ensemble() ,
    x , 0.0 , 1000.0 , dt );
```

Everything seems easy!

But how does lorenz\_ensemble look like?

# Ensemble of Lorenz systems

```
struct lorenz ensemble {
 size t N:
  state type beta;
 template< class State , class Deriv >
 void operator()(
    const State &x , Deriv &dxdt , value_type t ) const {
    thrust::for each (
      thrust::make_zip_iterator( thrust::make_tuple(
        x.begin() , x.begin()+N , x.begin()+2*N ,
        beta.begin() ,
        dxdt.begin(), dxdt.begin()+N, dxdt.begin()+2*N
      ) ) ,
      thrust::make_zip_iterator( thrust::make_tuple(
        x.begin()+N , x.begin()+2*N , x.begin()+3*N ,
        beta.end() ,
        dxdt.begin()+N, dxdt.begin()+2*N, dxdt.begin()+3*N
      ) ) ,
      lorenz_functor() );
};
```

### Ensemble of Lorenz systems

```
struct lorenz ensemble
 // ...
  struct lorenz functor
    template < class T > __host__ __device__
    void operator() ( T t ) const
     value type R = thrust::get< 3 >( t );
     value_type x = thrust::get< 0 >( t );
     value_type y = thrust::get< 1 >( t );
     value_type z = thrust::get< 2 >( t );
      thrust::qet < 4 > (t) = sigma * (y - x);
      thrust:: qet < 5 > (t) = R * x - y - x * z;
      thrust::get< 6 > ( t ) = -b * z + x * y;
```

# Advanced features - continued

#### Reference wrapper std::ref, boost::ref

#### The ODE and the observers are always passed by value

```
integrate_const{s,ode,x,0.0,1.0,dt,obs);
s.do_step(ode,x,t,dt);
```

#### Reference wrapper std::ref, boost::ref

#### The ODE and the observers are always passed by value

```
integrate_const(s,ode,x,0.0,1.0,dt,obs);
s.do_step(ode,x,t,dt);
```

#### Use std::ref or boost::ref to pass by reference

```
integrate_const{s, std::ref(ode), x, 0.0, 1.0, dt,
    std::ref(obs));
```

### Using Boost.Range

Use Boost.Range to integrate separate parts of the overall state

Example: Lyapunov exponents for the Lorenz system

#### Complete ODE = Lorenz system + Perturbation

- Calculate transients by solving only the Lorenz system (initialize x, y, z)
- Solve whole system (state + perturbations)

```
std::vector<double> x(6,0.0);
integrate(s,lorenz,make_pair(x.begin(),x.begin
          ()+3),0.0,10.0,dt);
integrate(s,lorenz_pert,x,10.0,1000.0,dt);
```

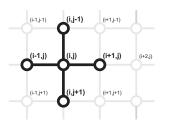
### ODEs with complex numbers

#### Discrete Nonlinear Schrödinger equation

$$\mathrm{i}\dot{\Psi}_k = arepsilon_k \Psi_k + V(\Psi_{k+1} + \Psi_{k-1}) - \gamma |\Psi_k|^2 \Psi_k \qquad , \quad \Psi_k \in \mathbb{C}$$

```
typedef std::vector<std::complex<double> > state_type;
struct dnls
  std::vector<double> eps;
 void operator() (const state type &x, state type &dxdt,
    double t) const
    const double V=0.5, gamma = 2.0;
    const complex<double> I(0.0,1.0);
    size t N = x.size();
    dxdt[0] = dxdt[N-1] = 0.0;
    for (size t i=1; i<N-1; ++i)
     dxdt[i] = -I * (eps[i]*x[i] + V*([x+1]+x[i-1])
        - gamma*norm(x[i])*x[i]);
};
```

#### Matrices as state types



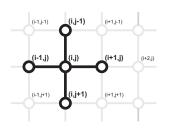
#### Example:

Two-dimensional phase lattice

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

```
typedef ublas::matrix<double> state_type1;
typedef mtl::dense2D<double> state_type2;
```

### Matrices as state types

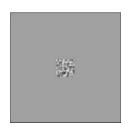


#### Example:

Two-dimensional phase lattice

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

typedef ublas::matrix<double> state\_type1;
typedef mtl::dense2D<double> state\_type2;







### Compile-time sequences and Boost.Units

$$\left(\begin{array}{c} \dot{X} \\ \dot{V} \end{array}\right) = \left(\begin{array}{c} V \\ f(X,V) \end{array}\right)$$

- x − length, dimension m
- v velocity, dimension ms<sup>-1</sup>
- a − acceleration, dimension ms<sup>-2</sup>

#### What else

ODEs on graphs



#### What else

ODEs on graphs

 Automatic memory management



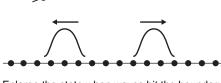
Enlarge the state when waves hit the boundary

#### What else

ODEs on graphs



 Automatic memory management



Enlarge the state when waves hit the boundary

Arbitrary precision types, GMPXX

- Introduction
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Technical details

4 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  $(\vec{x} = \vec{f}(\vec{x}, t))$ :

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

## 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

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

## 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:

```
Algebra algebra;

algebra.for_each3( x1 , x0 , F1 ,

Operations::scale_sum3( 1.0, b1*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:

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    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_each1( ... , 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.

۴n

## 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 >

#### $\verb|range_algebra \& default_operations| \textbf{work}| \textbf{ also with} \\$

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

#### What about

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

 $\longrightarrow$  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

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

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#### **Important**

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

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- 2 Tutorial

- Technical details
- Discussion