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CODE GENERATION FOR PARALLEL DIFFERENTIAL EQUATION SOLVERS

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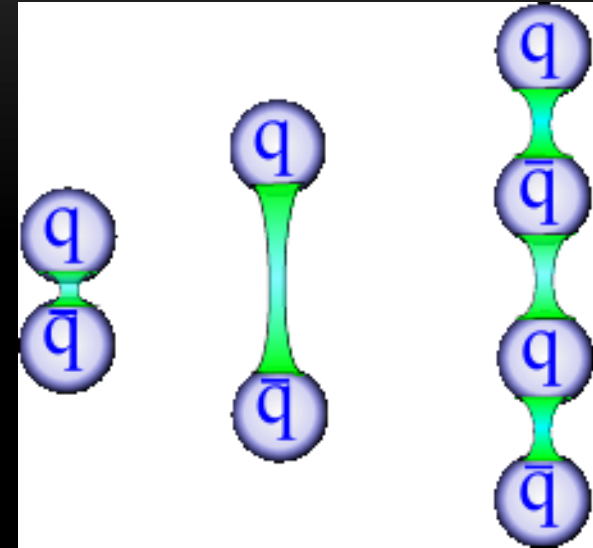
GPU Day 2015 – The Future of Many-Core Computing in Science
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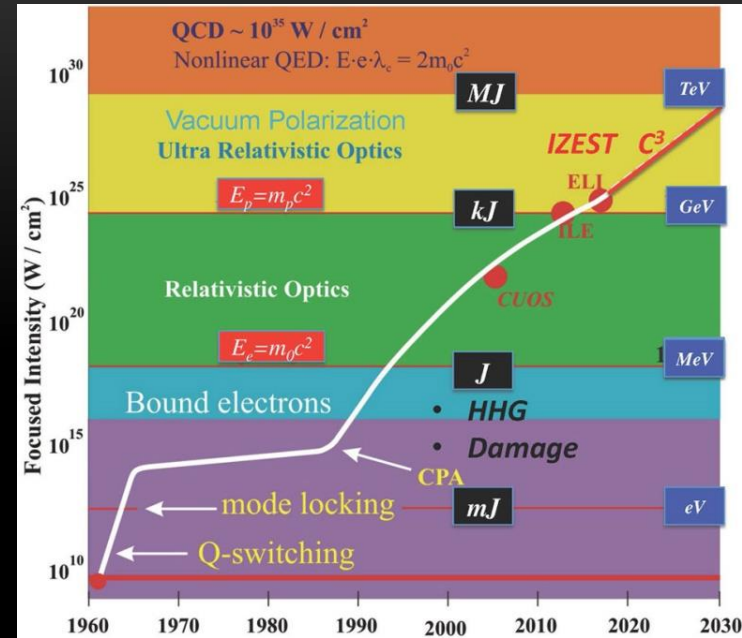
MOTIVATION

- Pair production from vacuum:
 - $q\bar{q}$ in heavy ion collisions:
 - Description of the early stages of collisions,
 - Formation and breaking of color strings,
 - Successful family of models,
 - Understanding of LHC data, particle spectras, etc.



MOTIVATION

- Pair production from vacuum:
 - e^+e^- in extreme strong laser fields:
 - Holy grail of QED
 - interesting non-linear phenomena in this regime
- Also possible near compact astrophysical objects

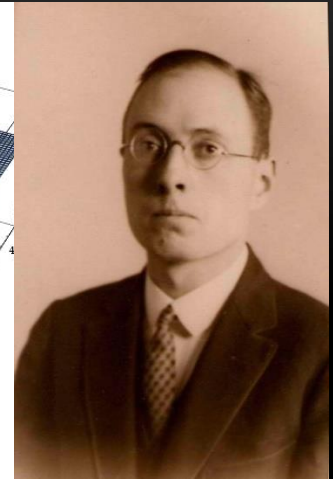
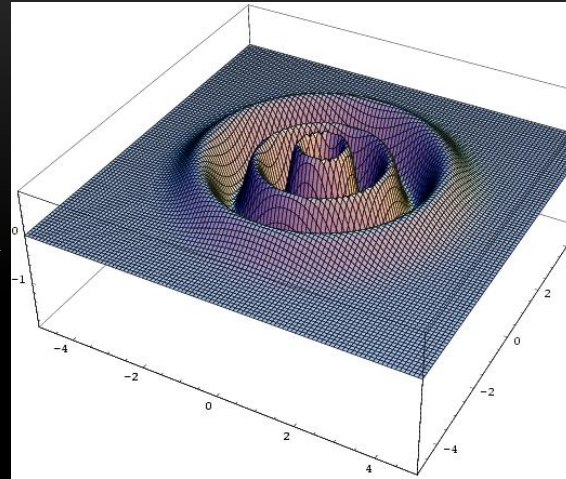


T. Tajima, G. Mourou



THEORETICAL MODEL

- **Wigner-functions**
- Quantum analogue of the classical one particle distribution function
- Definition:



Eugene Wigner

$$\hat{C}(\vec{x}, \vec{s}, t) = \exp\left(-iq \int_{-1/2}^{1/2} \vec{A}(\vec{x} + \lambda \vec{s}, t) \vec{s} d\lambda\right) \left[\Psi\left(\vec{x} + \frac{\vec{s}}{2}\right), \bar{\Psi}\left(\vec{x} - \frac{\vec{s}}{2}\right) \right]$$

$$W(\vec{x}, \vec{p}, t) = -\frac{1}{2} \int e^{-i\vec{p}\vec{s}} \langle 0 | \hat{C}(\vec{x}, \vec{s}, t) | 0 \rangle d\vec{s}$$

I. Bialynicki-Birula et al, Phys. Rev. **D44**, 1825-1835. (1991)

THEORETICAL MODEL

- Time evolution equations:

A Boltzmann like equations for the quantum one particle distribution function:
(QED: 3+3+1 dimension, 16 components, F. Hebenstreit et al, Phys. Rev. **D82** (2010) 105026.)

$$\begin{array}{rclcl}
 D_t \mathbb{S} & & - & 2\vec{P} \cdot \vec{t}_1 & = 0 \\
 D_t \mathbb{P} & & + & 2\vec{P} \cdot \vec{t}_2 & = 2m a_0 \\
 D_t \mathbb{V}_0 & + & \vec{D}_{\vec{x}} \cdot \vec{v} & & = 0 \\
 D_t a_0 & + & \vec{D}_{\vec{x}} \cdot \vec{a} & & = 2m \mathbb{P} \\
 D_t \vec{v} & + & \vec{D}_{\vec{x}} \mathbb{V}_0 & + & 2\vec{P} \times \vec{a} & = -2m \vec{t}_1 \\
 D_t \vec{a} & + & \vec{D}_{\vec{x}} a_0 & + & 2\vec{P} \times \vec{v} & = 0 \\
 D_t \vec{t}_1 & + & \vec{D}_{\vec{x}} \times \vec{t}_2 & + & 2\vec{P} \mathbb{S} & = 2m \mathbb{V} \\
 D_t \vec{t}_2 & - & \vec{D}_{\vec{x}} \times \vec{t}_1 & - & 2\vec{P} \mathbb{P} & = 0
 \end{array}$$

THEORETICAL MODEL

- The Wigner function based description can be extended to higher symmetries (**non-Abelian** case) too.

(Quark Wigner function evolution: A.V. Prozorkevich, S.A. Smolyansky, S.V. Ilyin)

- There will be more and more components
- The intermixing of components will be more complex
- If the **SU(N)** color matrices replaced by unity, the QED equations are recovered.

COMPUTATIONAL ASPECTS

Challenges of the evolution equations:

- Too many dimensions → Simplified configurations are obligatory
Magnetic field usually neglected
- Handling of the non-local differential operators is not trivial
- Extreme separation of time scales for realistic field parameters
- But most importantly:
extremely intense computational problem!

COMPUTATIONAL ASPECTS

Challenges of the evolution equations:

- Too many dimensions → Simplified configurations are obligatory
Magnetic field usually neglected
- Handling of the non-local differential operators is not trivial
- Extreme separation of time scales for realistic field parameters
- But most importantly:
extremely intense computational problem!

Fits well to GPUs!



WORKFLOW

- The expected workflow is:
 1. Formulation
(mathematical equations,
symbolic manipulation)
 2. Spectral expansion into
a dense matrix problem
(maybe inside a finite difference time
integrator)
 3. Efficient parallel
implementation/execution
- Possible tools:
 1. Maxima, Mathematica, ...
 2. Maxima, Mathematica, ...
with or exported to
Host high level language:
C++, Fortran...
 3. Parallel APIs & libraries:
OpenCL, CUDA, ...
BLAS implementations...

PROGRAMMING ASPECTS

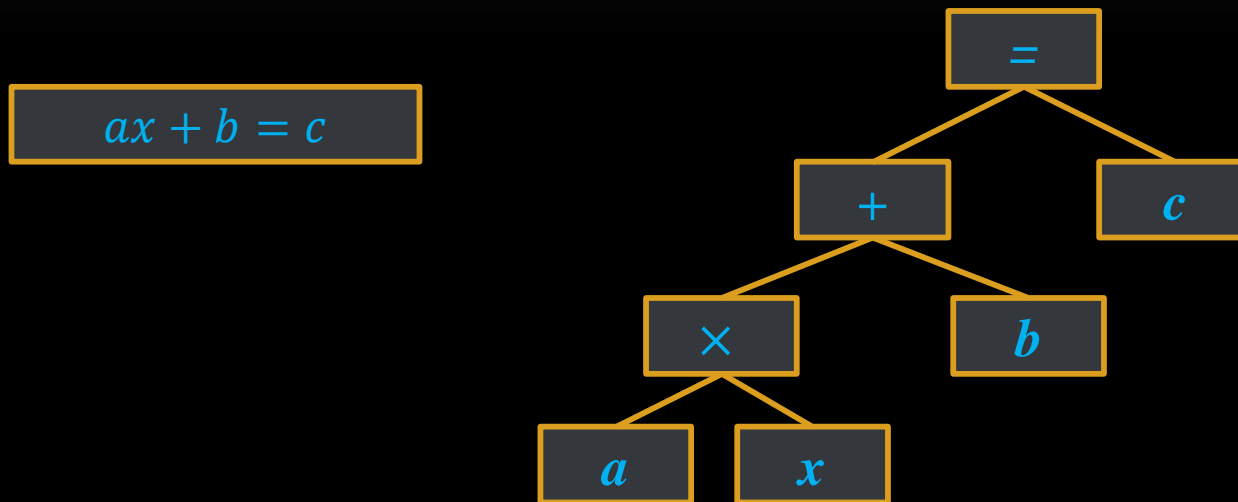
- We'd like to investigate many **electromagnetic vector field** configurations.
- But the equations may be drastically simplified by some choices, via symbolic algebraic manipulations.
- **One may not want to rewrite completely the equations each time, when the field is changed!**
- Less tools → less headache...
- Development time constrained!
- More **automatization** and **generic** solvers needed!
- We aim to generate the final specialized solver code **from the mathematical equations!**

PROPOSED ALTERNATIVE SOLUTION

- Run-time construction and manipulation of the **abstract syntax tree (AST)** from the **symbolic equations** and dynamic **code-generation!**
- Main advantages:
 - We can work with only one tool!
 - But, we can support many language back-ends and APIs
 - infer as much as possible from the symbolic equations!
 - **eliminate** many error sources and inconsistencies!

ABSTRACT SYNTAX TREES

- Mathematical formulas and equations can be represented by trees:



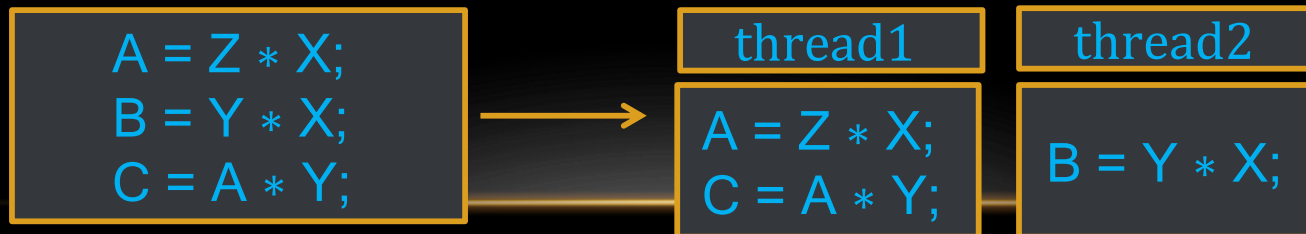
AST MANIPULATIONS

Since all of the needed constructs are trees the workflow can be seen as a series of transformations on the ASTs!

- **Symbolic (math) stage:**
 - simplifications ($0 \cdot a \rightarrow a$, $3a + 4a \rightarrow 7a$)
 - symbolic differentiation ($\frac{d\sin(x)}{dx} \rightarrow \cos(x)$)
 - series expansions ($f(x) \rightarrow \sum_{i=0}^n f_i \Phi_i(x)$)
check for argument sanity and rank mismatch
- **Programming stage:**
Infer types, further sanity checks

`a = dot(vector<2, double>(2., 9.), vector<2, double>(1., 0.))` → `a` is scalar double

Parallelization from data dependency (consider matrix operations):



AST MANIPULATIONS

At the symbolic stage a general model is specialized according to user defined constants and parameters and simplified symbolically.

Numerical solvers are just higher-order functions operating on the equations.

Example: Spectral Expansion (like Fourier, Chebyshev series)

1. Equation:

$$\frac{df(x)}{dx} = -af(x)$$

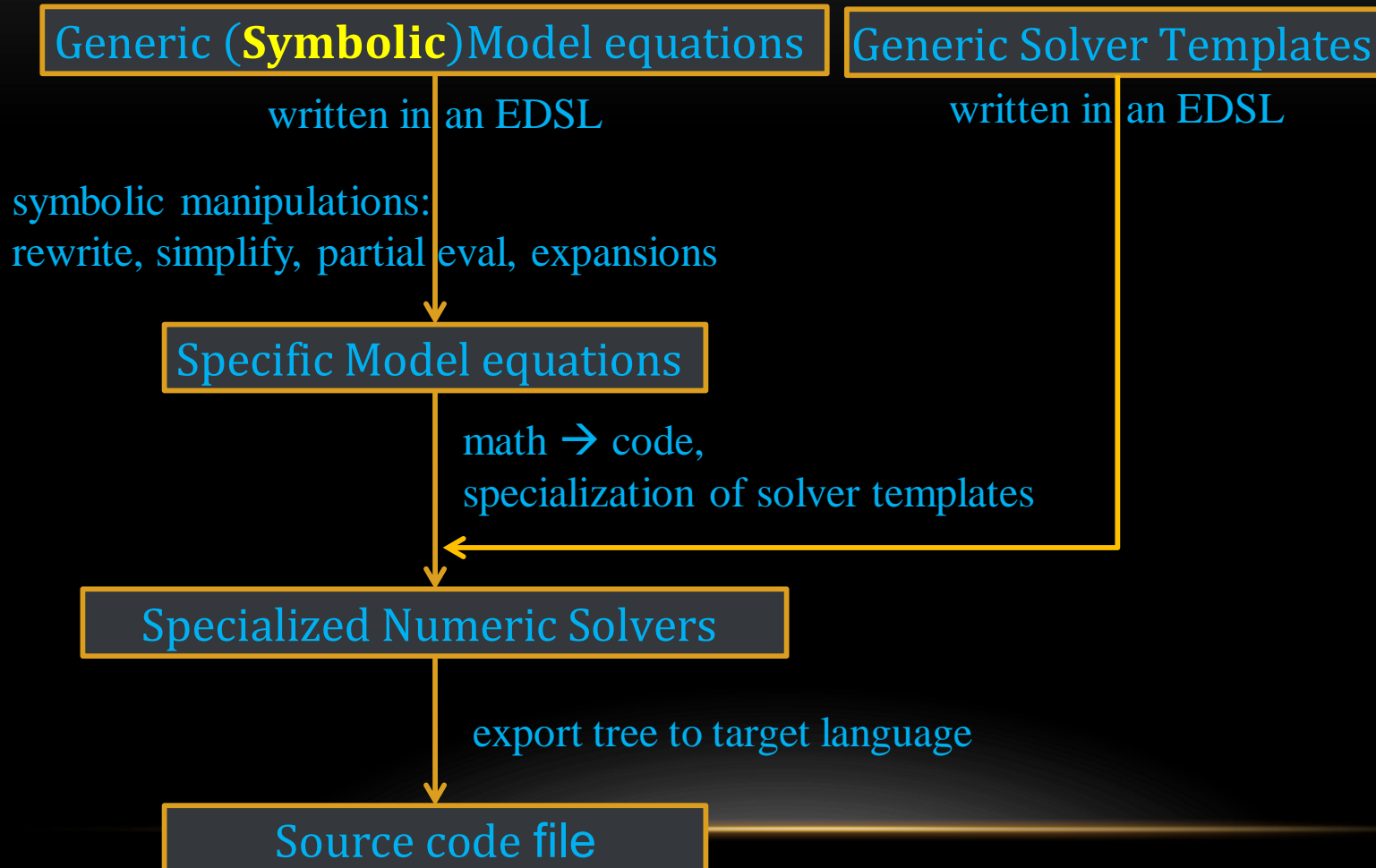
2. Expansion: $f(x) = \sum_{i=0}^N f_i \Phi_i(x)$

$$\sum_{i=0}^N f_i \frac{d\Phi_i(x)}{dx} = -a \sum_{i=0}^N f_i \Phi_i(x)$$

3. Differentiation: $\Phi_i(x) = \cos(iNx)$

$$-\sum_{i=0}^N f_i iN \sin(iNx) = -a \sum_{i=0}^N f_i \cos(iNx)$$

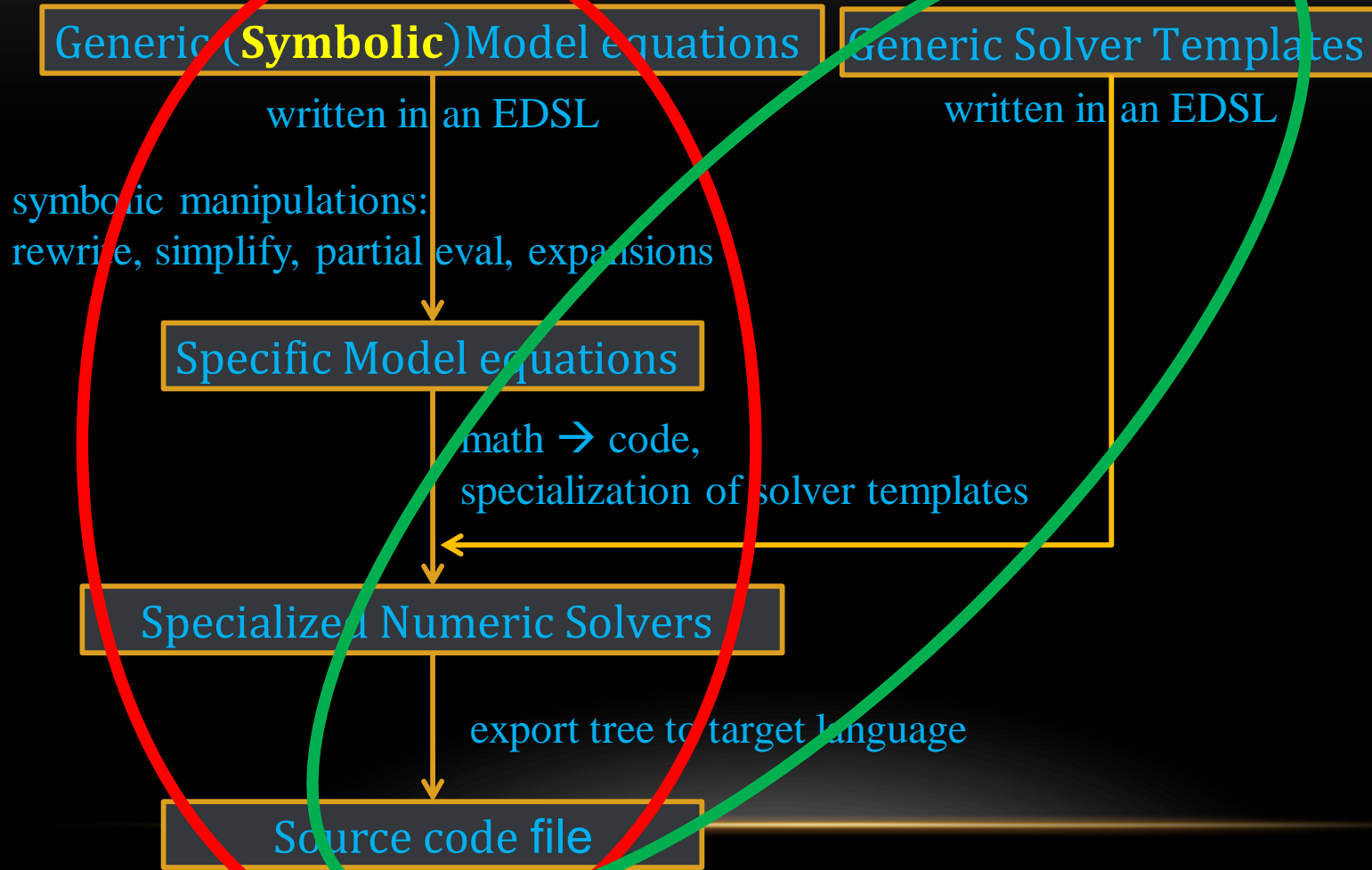
CODE-GENERATION FLOW



CODE-GENERATION FLOW

1.

2.



CURRENT STATUS

1. Math EDSL Project:

- Symbolic formulation and manipulation of equations, Spectral expansion
- Solution on the GPU with hand written solver
- Successfully tested on many PDEs:
Wave eq., Fokker-Planck eq., Vlasov eq., sQED Wigner

```
SymbolicDE DE;  
DE.DimensionsSymbols() << t << x;  
DE.UnknownSymbols() << f;  
DE.Equations() << diff(t, f) - D*diff(x, diff(x, f)) + v*diff(x, f);  
  
DE.Constants() << equate(t0, 0.0) << equate(D, 0.75) << equate(v, 1.2);  
  
DE.BoundaryConditions() << f(t0, x) - exp(-sq(x+2.)/(4*D*t0))/sqrt(4*pi*D*t0);  
  
DE.SpectralBases() << SpectralExpansion(L"Chebyshev", 42, 1.0, 5.0)  
    << SpectralExpansion(L"RationalChebyshev", 42, 0.0., 1.5);  
  
DE.ProcessAsFullSpectral();
```

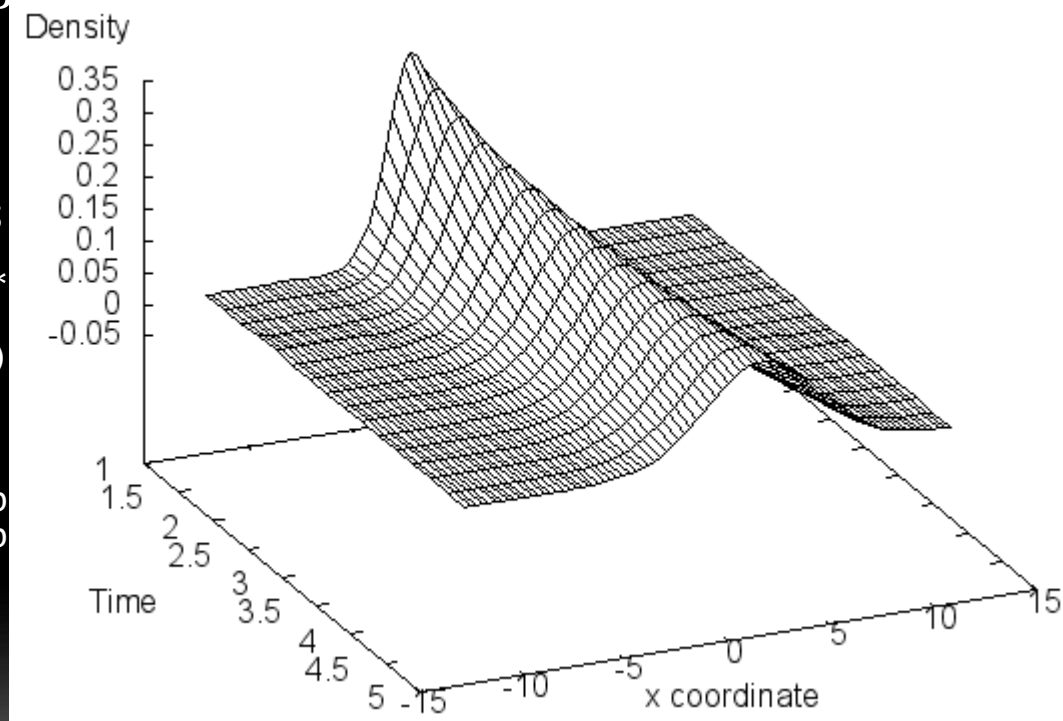
Syntax is subject to further optimization

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Wave eq., Fokker-Planck eq.,

```
SymbolicDE DE;  
DE.DimensionsSymbols() << t << x;  
DE.UnknownSymbols() << f;  
DE.Equations() << diff(t, f) - D*  
  
DE.Constants() << equate(t0, 0.0)  
  
DE.BoundaryConditions() << f(t0,  
  
DE.SpectralBases() << SpectralExp  
                << SpectralExp  
  
DE.ProcessAsFullSpectral();
```



CURRENT STATUS

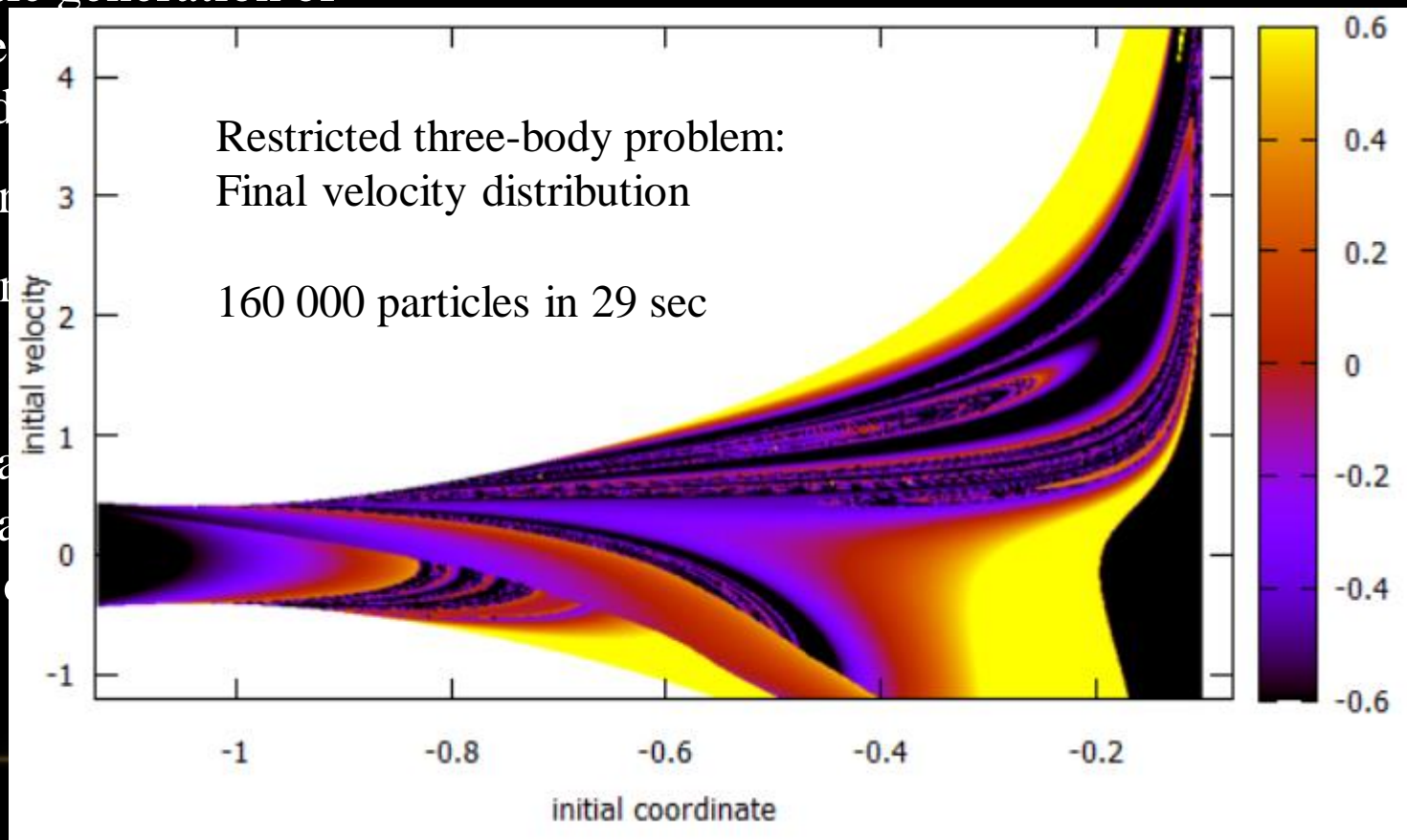
2. Solver Template EDSL Project:

- Expression of parallel programs with higher-order functions
- Automatic generation of host side C++ code and client side OpenCL code
- 8th order adaptive Runge-Kutta stepper implemented
- Tested on ODE systems, $\approx 30x$ faster than single thread CPU code.
- Current work is to cleanup and improve syntax of the EDSL in C++
- Current application:
 - collaboration with Gábor Drótos (Eötvös University, Budapest) to study chaotic scattering in the restricted three-body problem.
 - Collaboration with Alexandru Nicolin (Horia Hulubei National Institute, IFIN-HH) to model Bose-Einstein condensates.

CURRENT STATUS

2. Solver Template EDSL Project:

- Expression of parallel programs with higher-order functions
- Automatic generation of host side and client side code
- 8th order Runge-Kutta
- Tested on Intel Xeon Phi
- Current architecture and performance analysis in collaboration with Intel to study the impact of the architecture on the performance of the code



SUMMARY

- Efficient modeling of complicated equations in pair production demand some novel tools to handle special cases automatically while delivering high-performance computation e.g. on GPUs.
- Abstract mathematical models are easier to understand and automatically manipulate symbolically
- Same for abstract program code
- Dynamic code generation can combine these into efficient GPU simulations in a generic way

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THANK YOU

- Questions?

BACKUP SLIDES

THEORETICAL MODEL

- Hartee-type approximation: $\langle F^{\mu\nu} C \rangle \rightarrow \langle F^{\mu\nu} \rangle \langle C \rangle$
- No back reaction, no radiation corrections, etc.
- Compact form of evolution equation:

$$D_t W = -\frac{1}{2} \vec{D}_{\vec{x}} [\gamma^0 \vec{\gamma}, W] - im [\gamma^0, W] - i \vec{P} \{ \gamma^0 \vec{\gamma}, W \}$$

- The differential operators are non trivial operator series:

$$D_t = \partial_t + e \vec{E} \vec{\nabla}_p + \dots$$

- Expansion on 4x4 Dirac basis:

$$W(x, p, t) = \frac{1}{4} [1s + i\gamma_5 \mathbb{P} + \gamma^\mu \nabla_\mu + \gamma^\mu \gamma_5 a_\mu + \sigma^{\mu\nu} t_{\mu\nu}]$$

THEORETICAL MODEL

- In case of no magnetic field ($B=0$) and homogeneous electric field ($E(t)$):

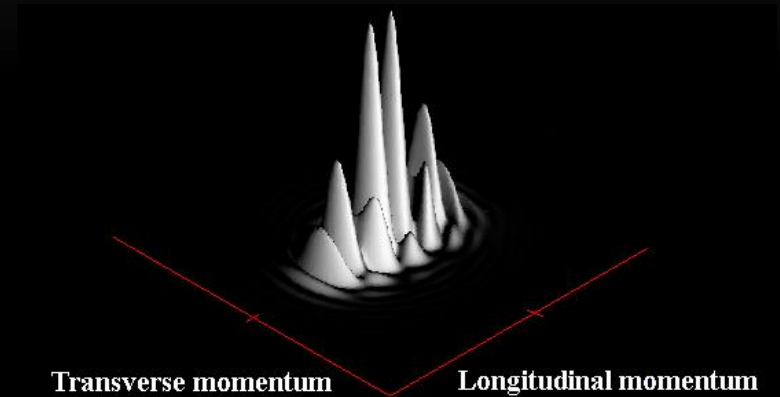
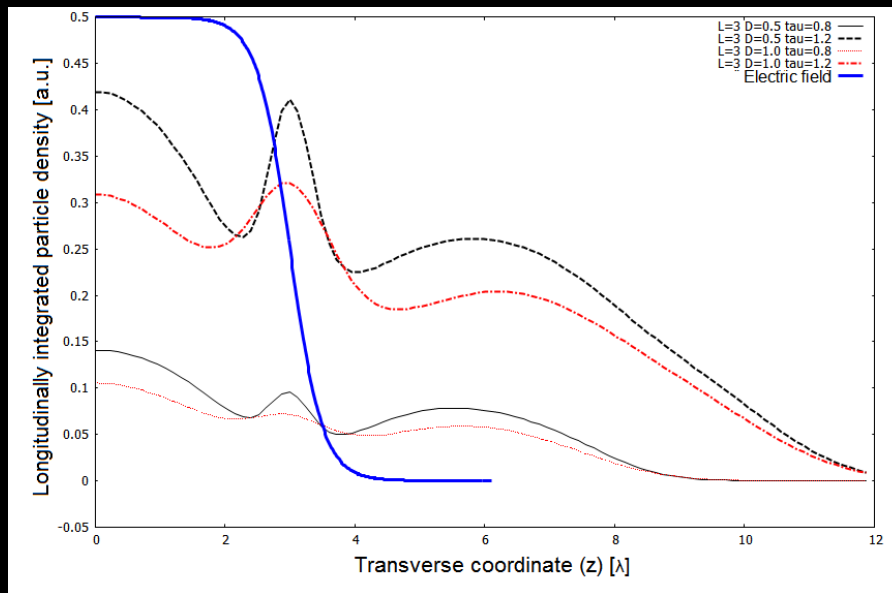
$$\frac{df}{dt} = \frac{eE\varepsilon_{\perp}}{\omega^2} v$$
$$\frac{dv}{dt} = \frac{1}{2} \frac{eE\varepsilon_{\perp}}{\omega^2} (1 - 2f) - 2\omega u$$
$$\frac{du}{dt} = 2\omega u$$

where:

$$\vec{p} = (\vec{q}_{\perp}, q_{\parallel} - eA(t))$$

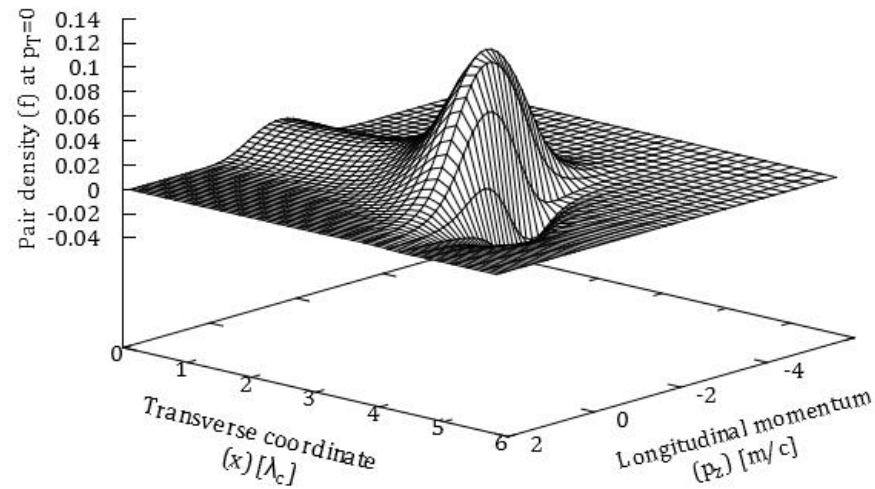
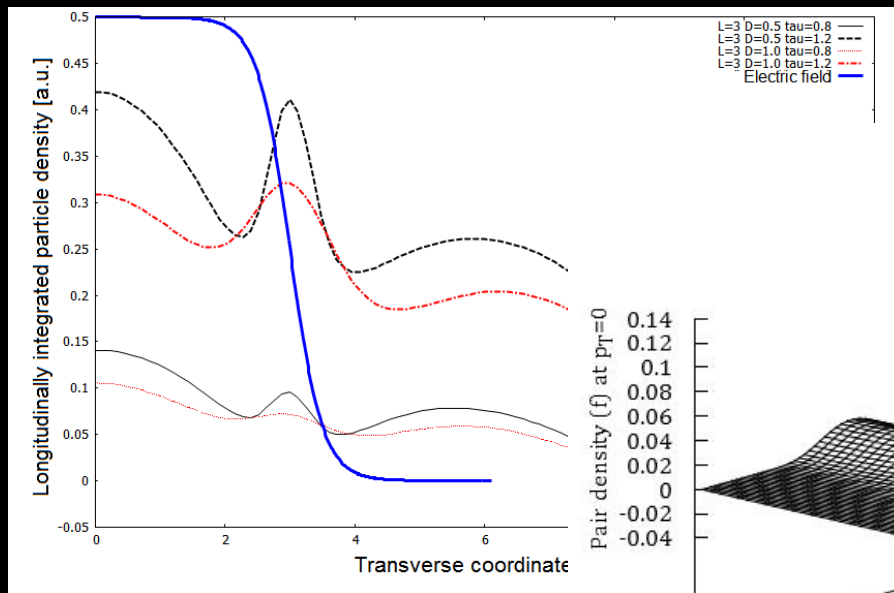
PHYSICS

- We can calculate asymptotic ($t \rightarrow \infty$) Wigner functions, so we have:
- Energy-, Momentum-, Mass-, Charge- and Spindensity.
- We can calculate particle spectras:



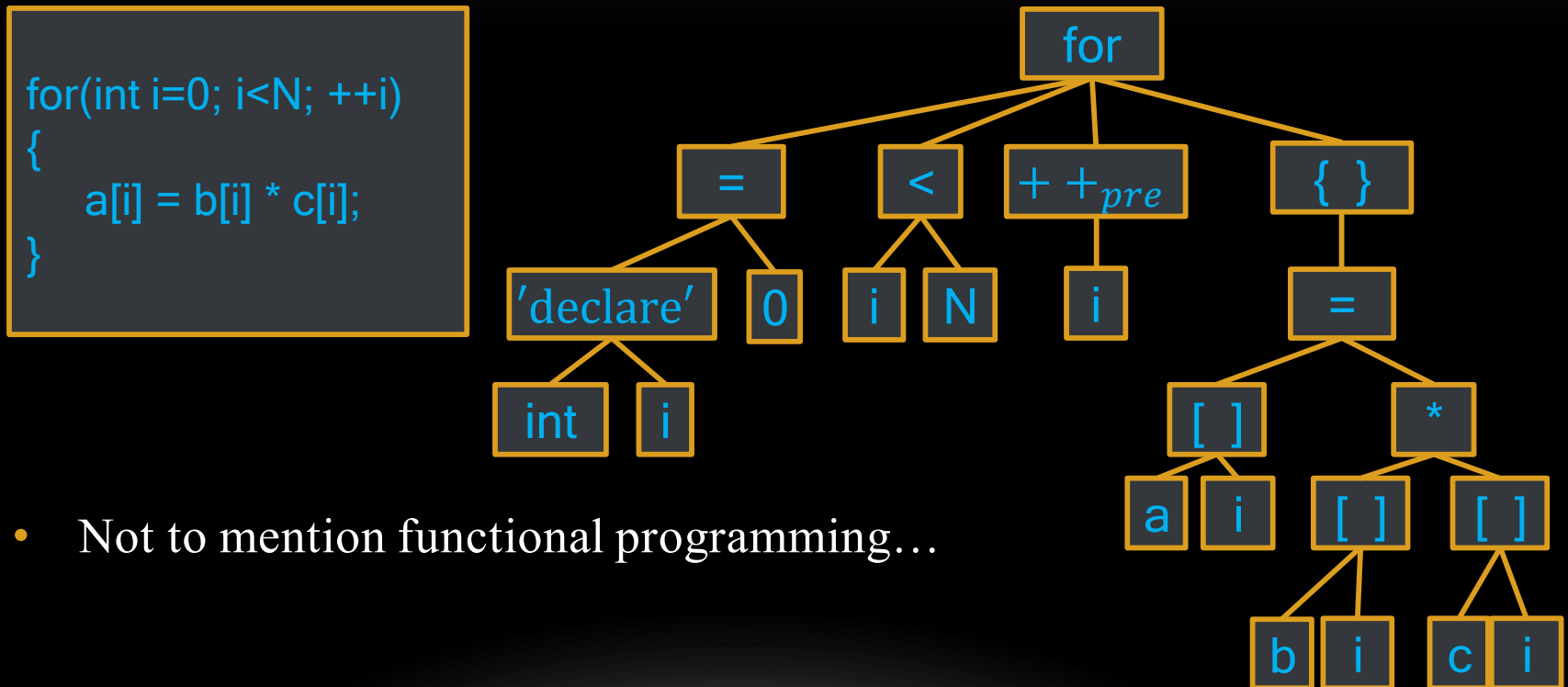
PHYSICS

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- Energy-, Momentum-, Mass-, Charge- and Spindensity.
- We can calculate particle spectras:



ABSTRACT SYNTAX TREES

- Imperative programming constructs can also be represented by trees:



- Not to mention functional programming...

SYMBOLIC \rightarrow PROGRAMMING CONVERSION

All Math objects are given a type deduced from the leaves and propagated upwards.

Function definitions, signatures constructed, defunctionalization applied.

One important construct: ParallelFunction created from vector, matrix operations!

```
f1(a,  $\vec{x}$ ,  $\vec{y}$ ):= $a\vec{x} + \vec{y}$ 
```

```
 $\vec{z} = f1(a, \vec{x}, \vec{y})$ 
```

```
void f1(range1 i, double a, double* z,  
        double* x, double* y)
```

```
{
```

```
    z[i] = a * x[i] + y[i];
```

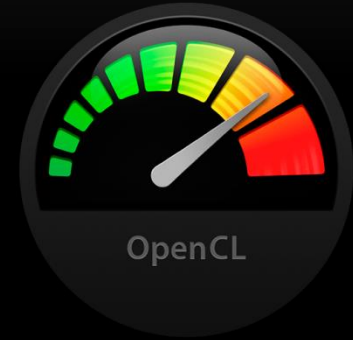
```
}
```

Calls are generated as:

```
ParallelCall( f1, RangeOf(z), a, z, x, y);
```

FINAL CODE-GENERATION

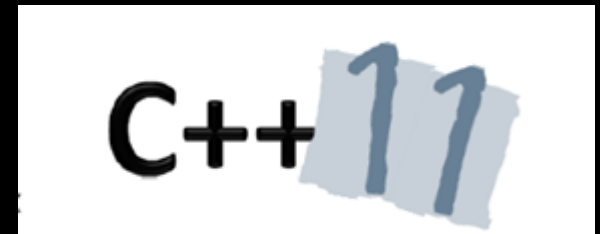
- When all the conversions are ready the program tree is traversed and all the branch operators are converted to their textual equivalents in the selected languages.
- Currently C++ / C / OpenCL export is considered.



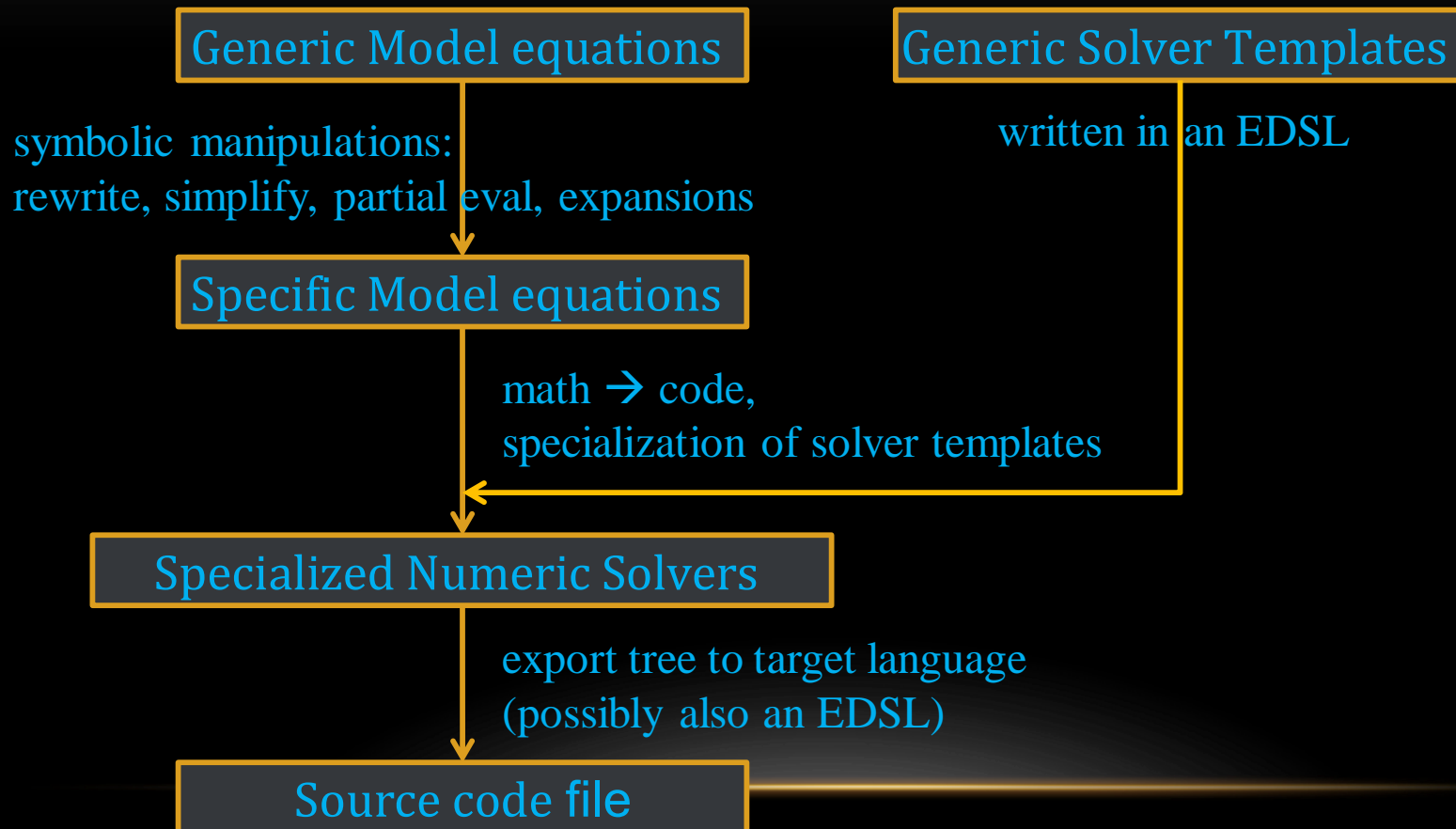
Why the C family?

Largest common set of features supported by the compute and rendering APIs:

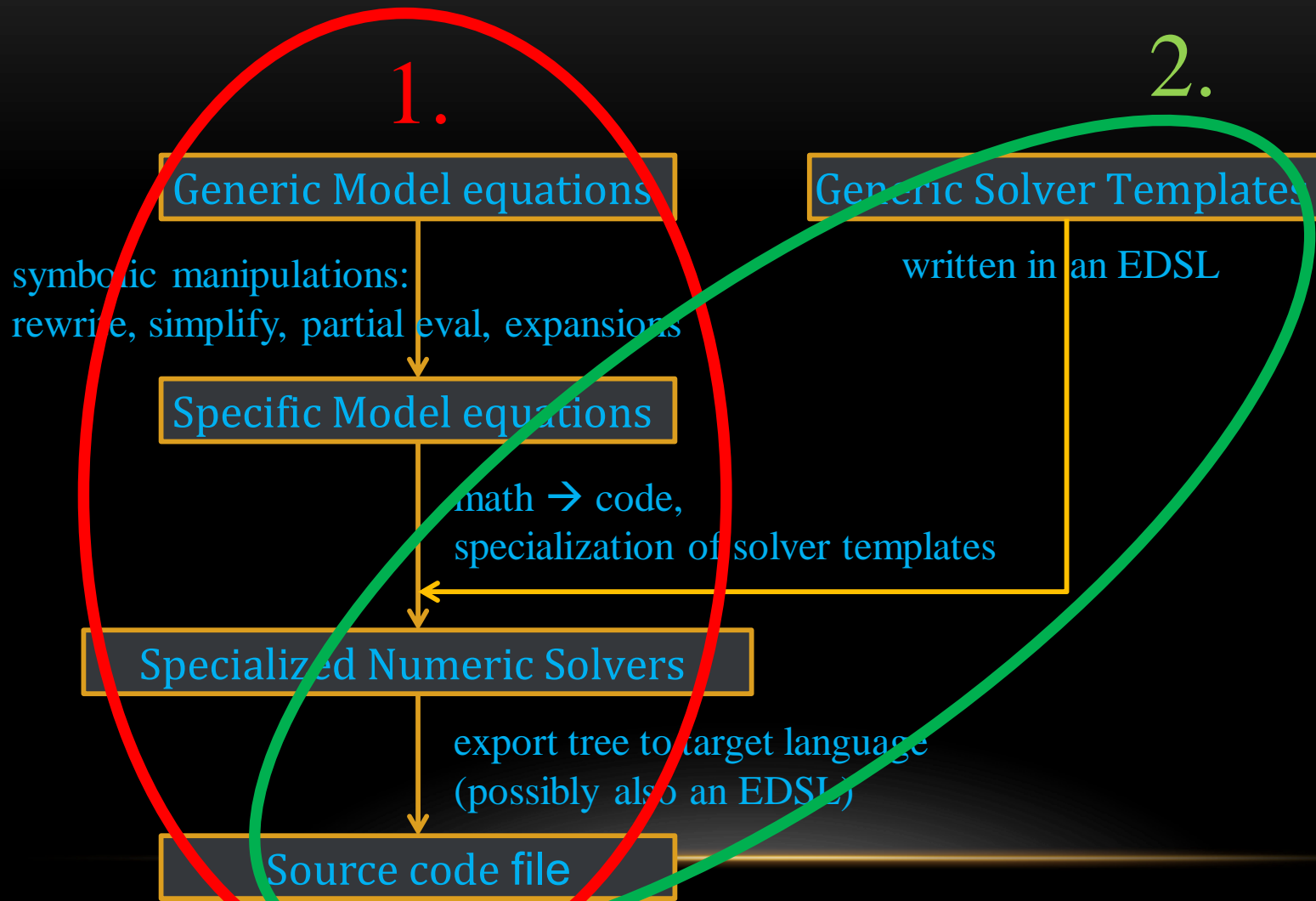
OpenCL kernel C, OpenGL GLSL,
DirectCompute, DirectX HLSL



CODE-GENERATION FLOW



CODE-GENERATION FLOW

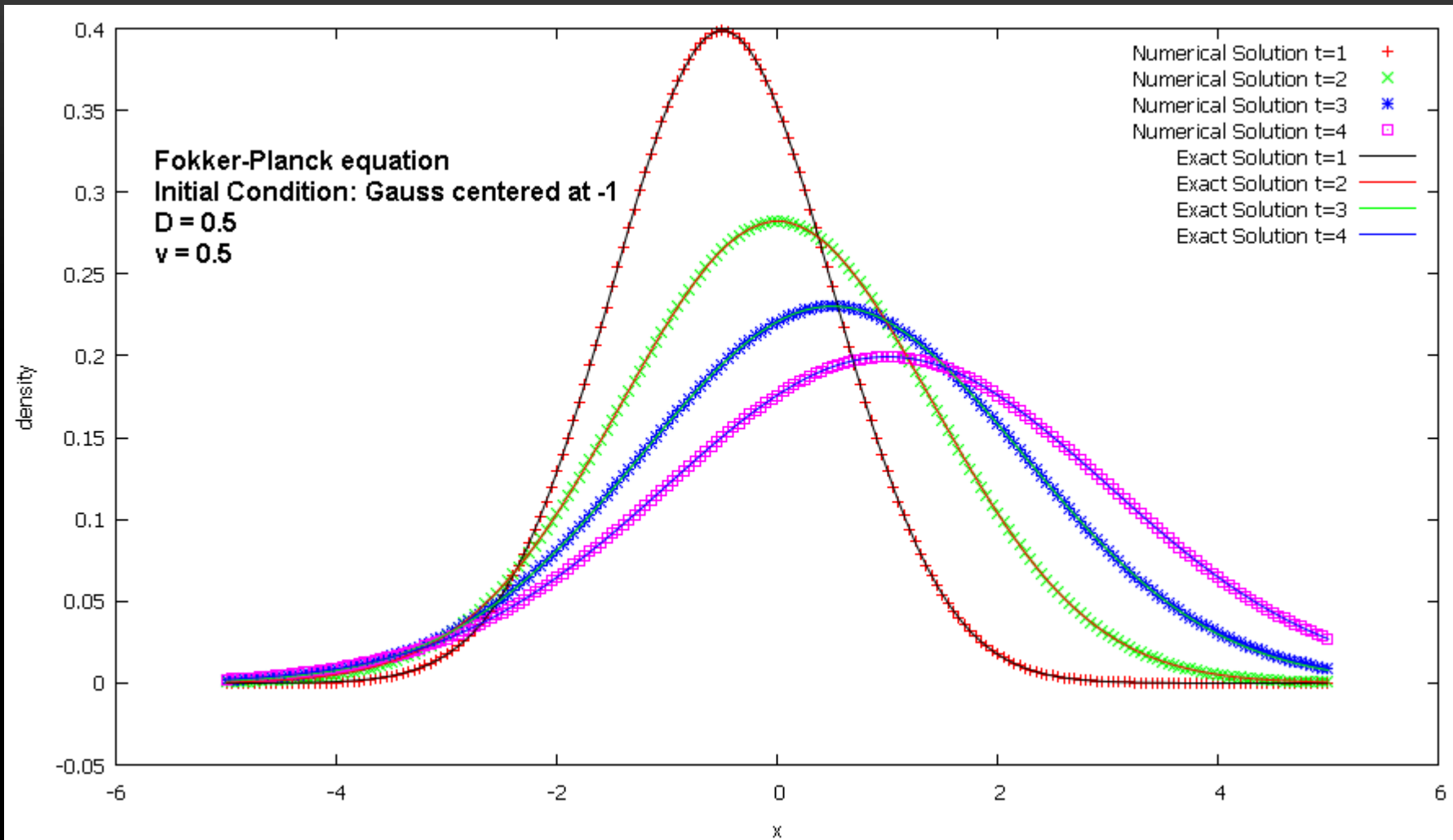


```

1 #include <Phys/DifferentialEquations.h>
2
3 void FokkerPlanckEquation()
4 {
5     SymbolicDE DE;
6
7     MathExpr t(L"t", 1, 1);
8     MathExpr x(L"x", 1, 1);
9     MathExpr f(L"f", 1, 1);
10    MathExpr D(L"D", 1, 1);
11    MathExpr v(L"v", 1, 1);
12    MathExpr t0(L"t0", 1, 1);
13    MathExpr pi(L"PI");
14
15    DE.DimensionSymbols() << t << x;
16    DE.UnknownSymbols() << f;
17    DE.Equations() << diff(t, f) - diff(x, diff(x, D(x)*f)) + diff(x, v*f);
18    DE.Constants() << equate(t0, 0.0);
19    DE.Functions() << equate(D, 0.5) << equate(v, 0.5);
20
21    DE.BoundaryConditions()
22        << f(t0, x) - exp(-sq(x+v*t0)/(D*4*t0))/sqrt(pi*4.0*t0*D);
23
24
25    DE.SpectralBases() << SpectralExpansion(L"RationalChebyshev", 48, 0.0, 1.0)
26        << SpectralExpansion(L"RationalChebyshev", 48, 0.0, 1.5);
27
28    DE.ProcessAsFullSpectral();
29
30    arr<double> ev; ev << 1.0 << 0.0;
31    DE.SampleSolutionToFile1(L"out.txt", -5.0, 5.0, 0.05, 1, ev );
32    arr<double> ev2; ev2 << 2.0 << 0.0;
33    DE.SampleSolutionToFile1(L"out2.txt", -5.0, 5.0, 0.05, 1, ev2 );
34    arr<double> ev3; ev3 << 3.0 << 0.0;
35    DE.SampleSolutionToFile1(L"out3.txt", -5.0, 5.0, 0.05, 1, ev3 );
36    arr<double> ev4; ev4 << 4.0 << 0.0;
37    DE.SampleSolutionToFile1(L"out4.txt", -5.0, 5.0, 0.05, 1, ev4 );
38
39    DE.SampleSolutionToFile2(L"fp.txt", -20.0, 20.0, 0.5, 0, -10.0, 10.0, 0.25, 1, ev4 );
40 }

```

DEMONSTRATION



```

1 #include "Computation.h"
2 #include "odes.h"
3 #include "Ex2.h"
4
5 struct RKState{ double x, v, t; double& operator[]( int i ){ return ((double*)&x)[i]; } };
6
7 void RK8Test()
8 {
9     using namespace Metaprogramming;
10
11     MetaProgram p;
12     ID(a); //identifier for user input
13     {
14         ID(x); ID(v); ID(t); ID(s); ID(i); ID(ss); //identifiers
15
16         Type Num("double"), State("State"), Int("int"); //type identifiers
17
18         //State struct
19         p |= decllist( Num|x, Num|v, Num|t ) | State;
20
21         //RHS of DE
22         p |= signature(State, State) | Id("rhs") = $(s, { !(State|ss), ss[~x] = s[~v], ss[~v] = -2.0*s[~x], rt(ss) });
23
24         //Indexer function for State
25         p |= signature(Num, State, Int) | Id("indexer") = $( ids(s,i), { rt( Select( i==0, s[~x], s[~v] ) ) });
26
27         //Higher order solver function definition imported:
28         p |= getRK8(Id("indexer"));
29
30         //Main entry point and solver invoke (translates to kernel call)
31         p |= signature(Type::Void(), vec(Num) ) | Id("main") = $(a, async_block( !Id("rk8")(domof(a), a, Id("rhs"), 2, Id("indexer")) ) );
32     }
33
34     Namespace ns;
35     au state = CreateBuffer<RKState>(200); //user buffer in CPU RAM
36     for( int i=0; i<state.ext[0]; i++ ){ state[i].x = 0.0; state[i].v = 2.0*i; state[i].t = 0.0; }
37
38     ns.CreateBuffer(a, state ); //Bind to identifier
39     ns.AddCode(p); //Compile metaprogram
40     ns.exec( a ); //Compile and Launch with identifier as parameter
41     ns.ReadBuffer("a"); //Read back to user buffer
42 }

```

DEMONSTRATION