

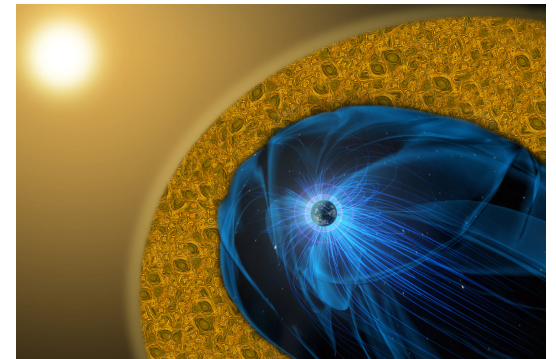
Advantages and pitfalls of OpenCL in computational physics



K. Ostaszewski, P. Heinisch, H. Ranocha

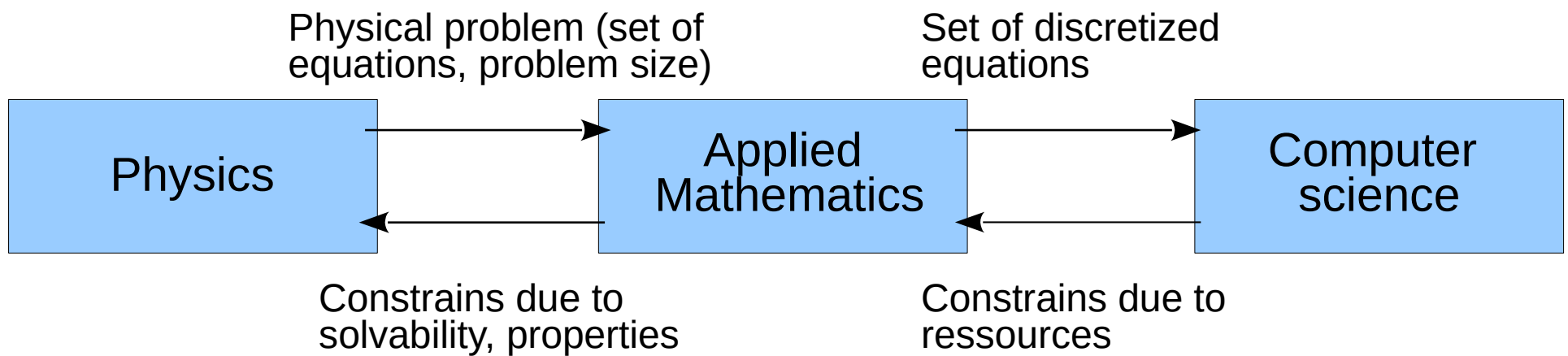
Computational Physics

- Very few physical problems can be solved analytically due to
 - Complexity,
 - Lack of algebraic solvability.
- Numerical approximations are required.
- Fields of application:
 - Plasmaphysics (Fusion, Astrophysics, industrial plasmas),
 - Weather prediction,
 - Solid state physics, etc.



Computational Physics

- Overlap of physics, applied mathematics and computer science.



- Different fields constrain eather other.

Motivation

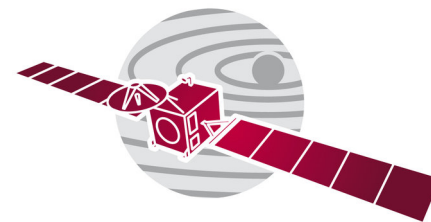
- Legacy Plasmacode (A.I.K.E.F) used for different NASA/ESA missions.



cassini-huygens



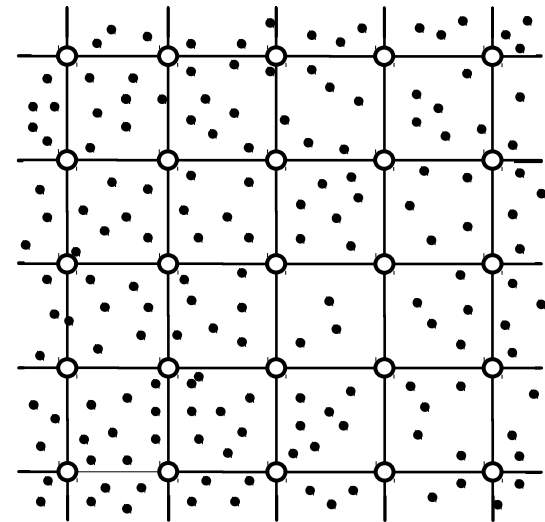
bepicolombo



- Parallelized with MPI (CPU based).
- Limits in scalability and resources reached.

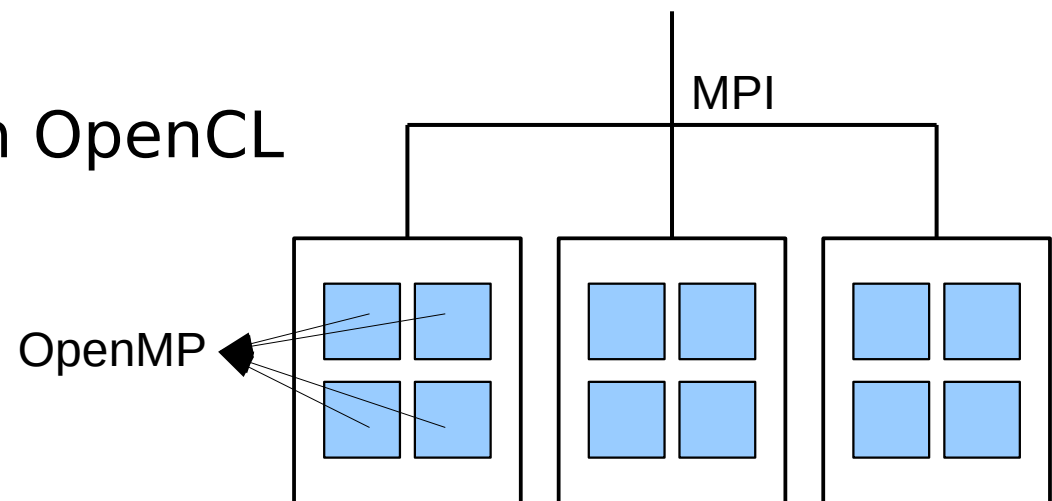
Plasmaphysics

- Different models to describe a plasma: Fluid-Model, Hybrid-Model, Particle-in-Cell (PiC).
- Need to describe particles and electromagnetic fields.
- Hybrid-Model includes the following mathematical problems:
 - Systems of linear equations,
 - Ordinary/partial differential equations,
 - N-Body interactions.



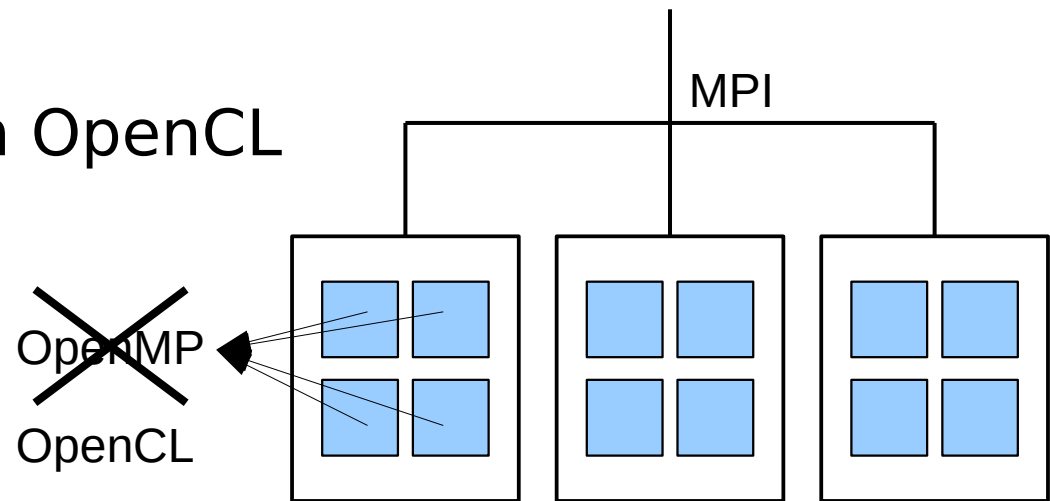
Simulation Approach

- Usual approach is to combine a global inter-node based MPI parallelization with a local OpenMP parallelization.
- Developed for deployment on clusters (HLRN,...)
- Substitute OpenMP with OpenCL



Simulation Approach

- Usual approach is to combine a global inter-node based MPI parallelization with a local OpenMP parallelization.
- Developed for deployment on clusters (HLRN,...)
- Substitute OpenMP with OpenCL



Advantages and pitfalls

Advantages of OpenCL

- Deployment on heterogenous systems (portability),
- Runtime advantage by using GPUs,
- Easy testing/changing of numerical submodules (Python, Oclgrind, MatCL),

Advantages and pitfalls

Advantages of OpenCL

- Deployment on heterogenous systems (portability),

- Run

- Easy (M

```
180
181   convec.x = d_field_b_old[idx].x*(d_field_u[increment_x(idx,1)].x - d_field_u[increment_x(idx,-1)].x)/((2.0f*(REAL)DX))
182             +d_field_b_old[idx].y*(d_field_u[increment_y(idx,1)].x - d_field_u[increment_y(idx,-1)].x)/((2.0f*(REAL)DY))
183             -0.5*d_field_b_old[idx].x*((d_field_u[increment_x(idx,1)].x - d_field_u[increment_x(idx,-1)].x)/(2.0f*(REAL)DX)
184             +(d_field_u[increment_y(idx,1)].y - d_field_u[increment_y(idx,-1)].y)/((2.0f*(REAL)DY)))
185             -0.5*(d_field_u[idx].x*(d_field_b_old[increment_x(idx,1)].x - d_field_b_old[increment_x(idx,-1)].x)/(2.0f*(REAL)DX)
186             +d_field_u[idx].y*(d_field_b_old[increment_y(idx,1)].x - d_field_b_old[increment_y(idx,-1)].x)/(2.0f*(REAL)DY))
187             -0.5*((d_field_u[increment_x(idx,1)].x*d_field_b_old[increment_x(idx,1)].x
188             -d_field_u[increment_x(idx,-1)].x*d_field_b_old[increment_x(idx,-1)].x)/(2.0f*(REAL)DX)
189             +(d_field_u[increment_y(idx,1)].y*d_field_b_old[increment_y(idx,1)].x
190             -d_field_u[increment_y(idx,-1)].y*d_field_b_old[increment_y(idx,-1)].x)/(2.0f*(REAL)DY));
191
192   convec.y = d_field_b_old[idx].x*(d_field_u[increment_x(idx,1)].y - d_field_u[increment_x(idx,-1)].y)/((2.0f*(REAL)DX))
193             +d_field_b_old[idx].y*(d_field_u[increment_y(idx,1)].y - d_field_u[increment_y(idx,-1)].y)/((2.0f*(REAL)DY))
194             -0.5*d_field_b_old[idx].y*((d_field_u[increment_x(idx,1)].x - d_field_u[increment_x(idx,-1)].x)/(2.0f*(REAL)DX)
195             +(d_field_u[increment_y(idx,1)].y - d_field_u[increment_y(idx,-1)].y)/((2.0f*(REAL)DY)))
196             -0.5*(d_field_u[idx].x*(d_field_b_old[increment_x(idx,1)].y - d_field_b_old[increment_x(idx,-1)].y)/(2.0f*(REAL)DX)
197             +d_field_u[idx].y*(d_field_b_old[increment_y(idx,1)].y - d_field_b_old[increment_y(idx,-1)].y)/(2.0f*(REAL)DY))
198             -0.5*((d_field_u[increment_x(idx,1)].x*d_field_b_old[increment_x(idx,1)].y
199             -d_field_u[increment_x(idx,-1)].x*d_field_b_old[increment_x(idx,-1)].y)/(2.0f*(REAL)DX)
200             +(d_field_u[increment_y(idx,1)].y*d_field_b_old[increment_y(idx,1)].y
201             -d_field_u[increment_y(idx,-1)].y*d_field_b_old[increment_y(idx,-1)].y)/(2.0f*(REAL)DY));
202
```

Advantages and pitfalls

Pitfalls of OpenCL

- Copy overhead serious bottleneck,
- Lack of debugging capability (e.g. no buffer overflow check),
- Documentation/examples lacking.

Partial Differential Equations

- Most simulation codes involve solving of differential equations on a discretized grid:

$$\nabla \times \vec{E} = -\frac{\partial \vec{B}}{\partial t} \qquad \nabla \times \vec{B} = \mu_0 \vec{j} + \frac{1}{c^2} \frac{\partial \vec{E}}{\partial t}$$

- Discretization schema influences physical properties of solution.
 - Trade of between accuracy and computational cost!

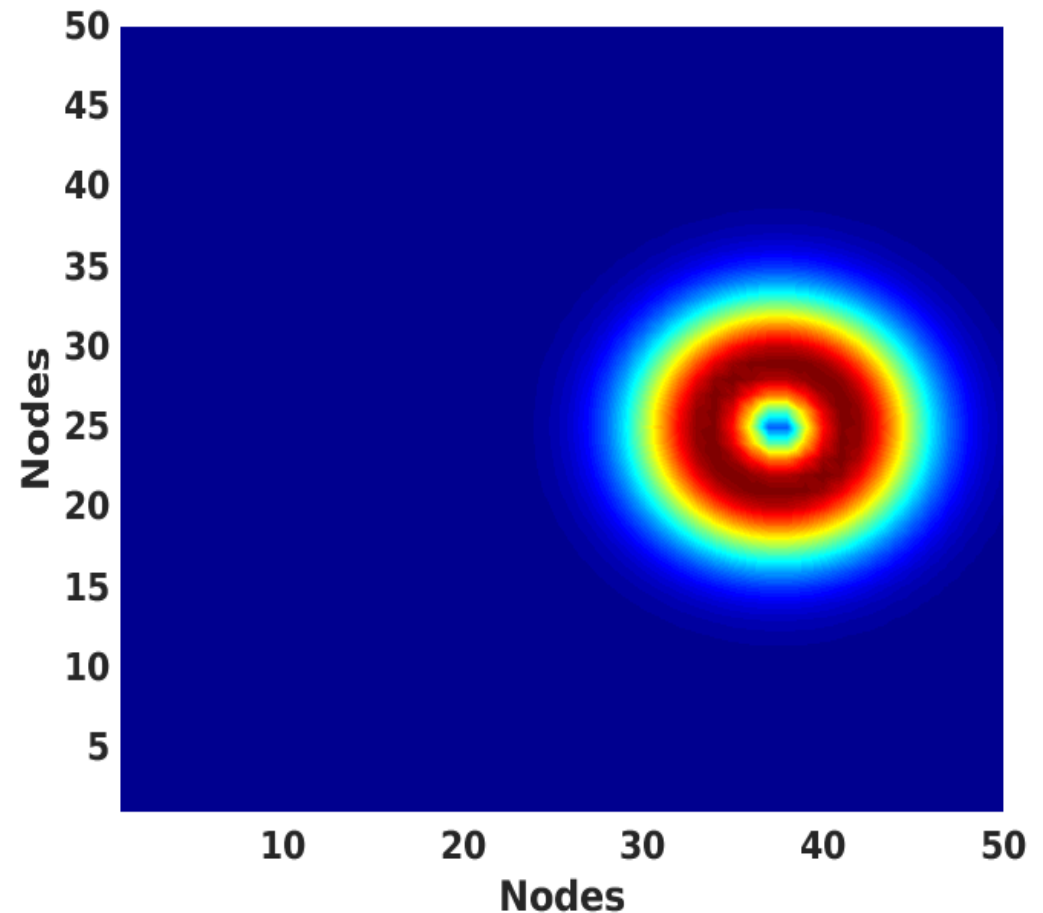
Example: Partial Differential Equations

- Example: Frozen-in-Theorem

$$\frac{\partial \vec{B}}{\partial t} = \nabla \times (\vec{u} \times \vec{B})$$

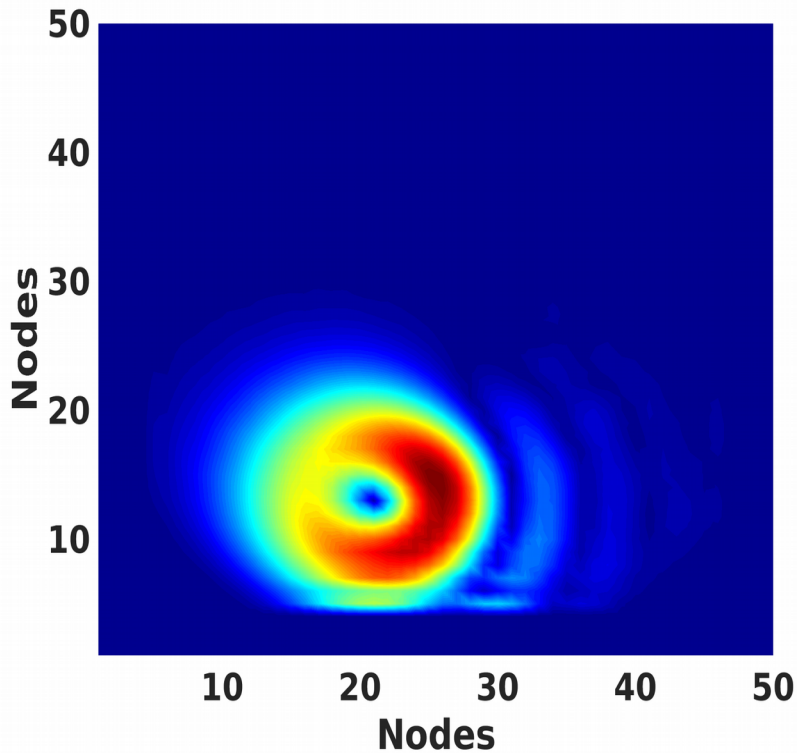
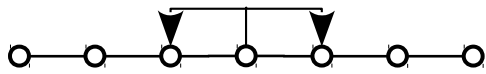
B: Magnetic field
u: Fluid velocity

- Solution will rotate around middle of the box.

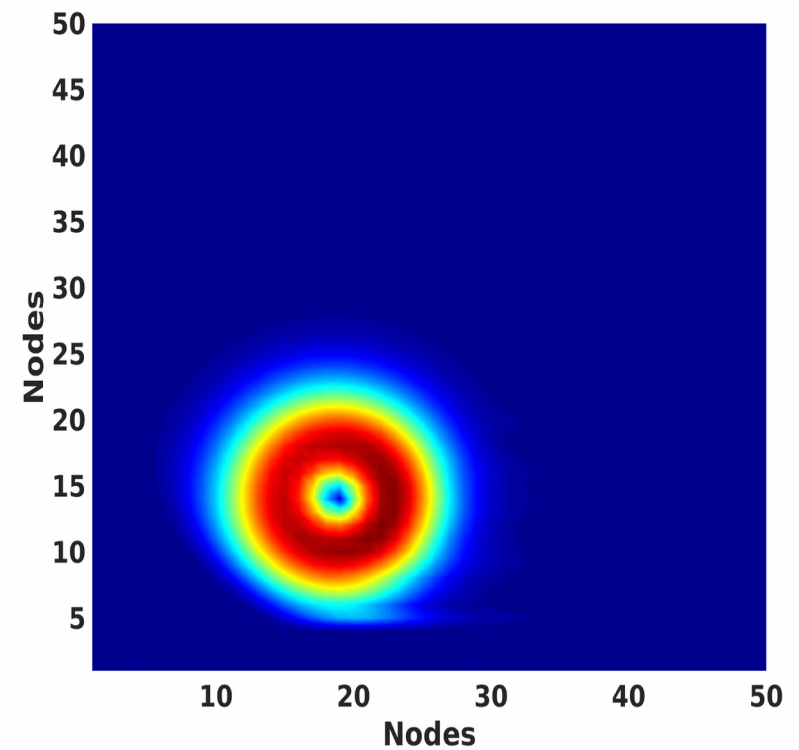
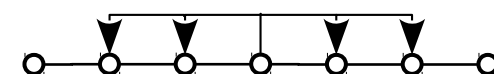


Example: Partial Differential Equations

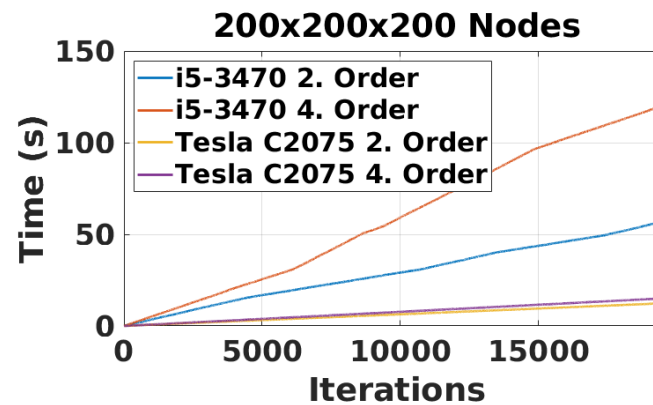
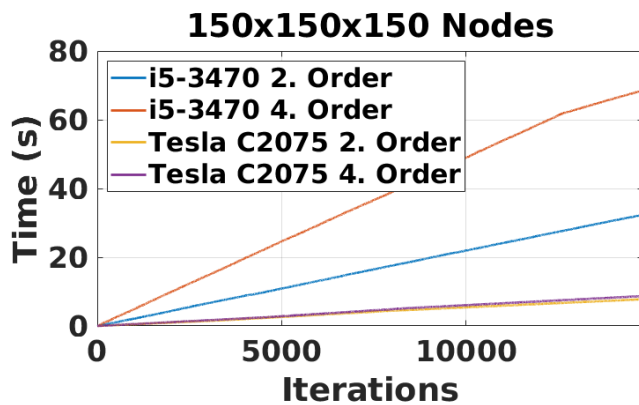
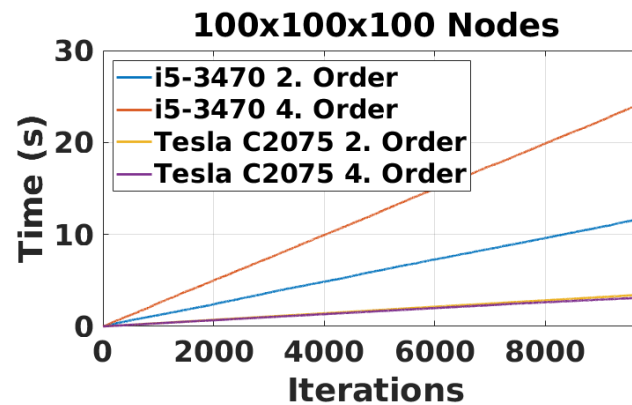
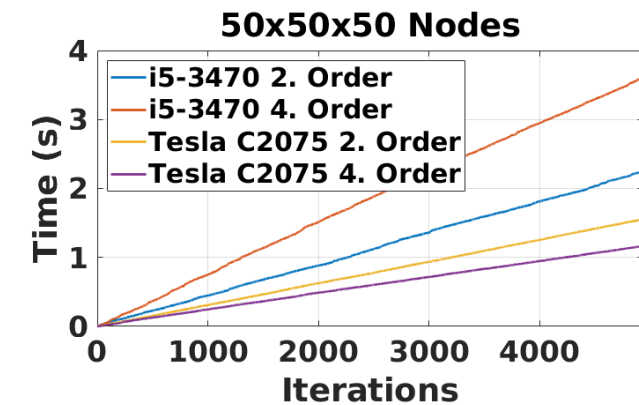
- Solver 2nd Order



- Solver 4th Order



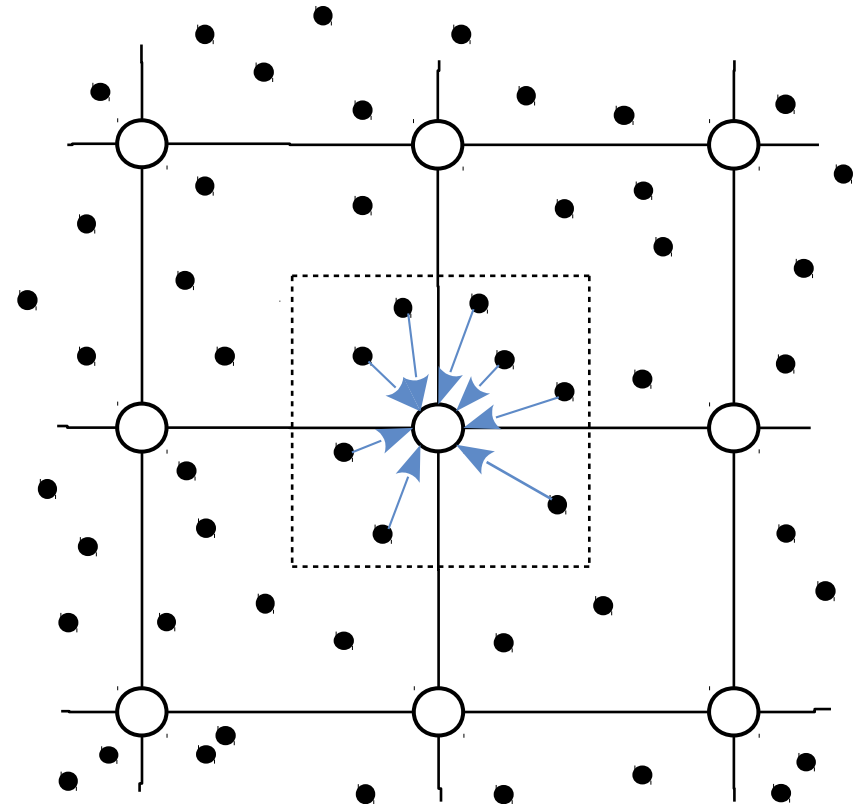
Example: Partial Differential Equations



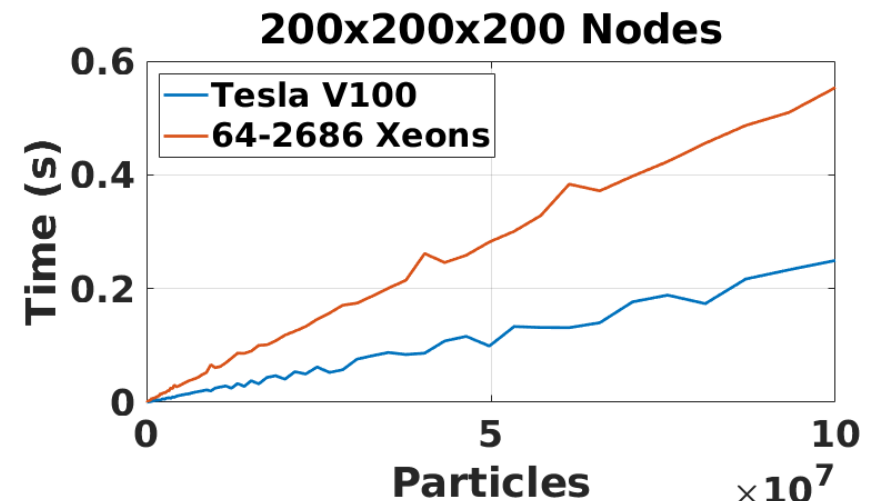
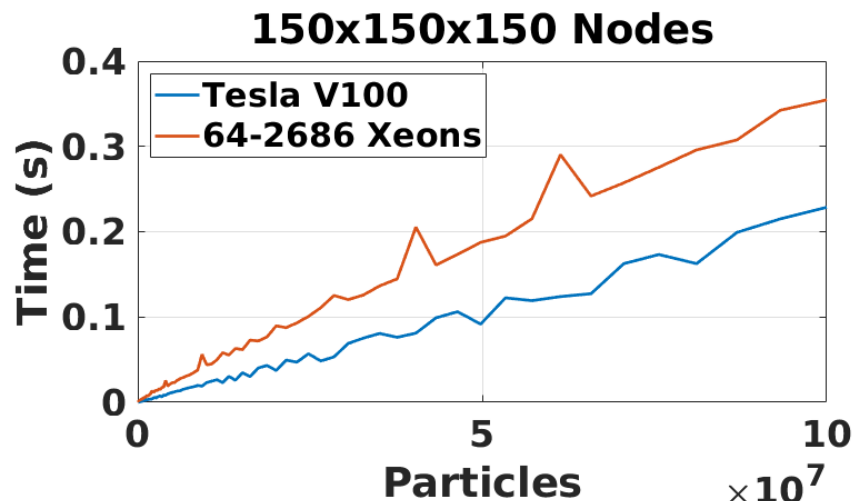
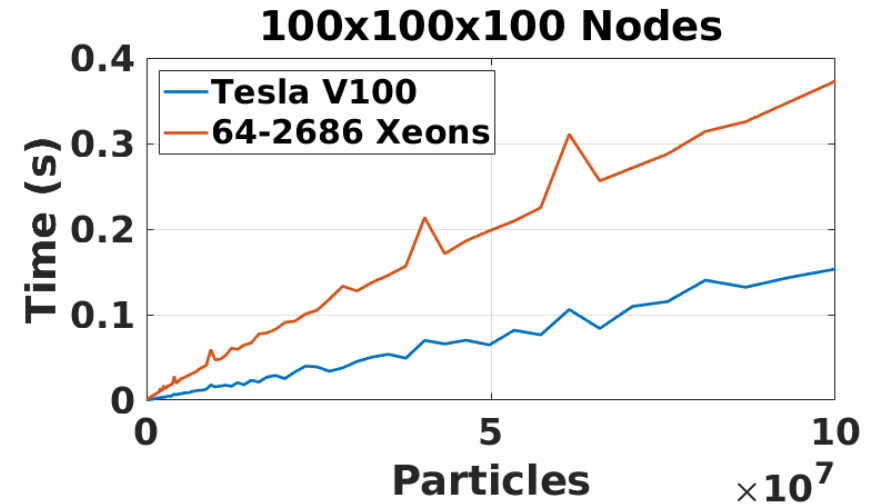
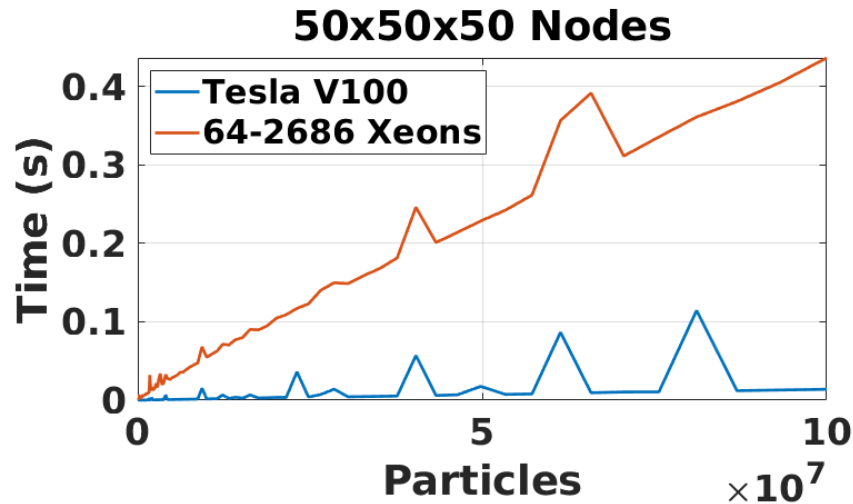
Number of Nodes	Performance penalty
	GPU CPU
50x50x50	0.78 1.63
100x100x100	0.95 2.04
150x150x150	1.11 2.23
200x200x200	1.21 1.89

Particle to Grid Reduction

- Sum up all particle on next node.
- Needed to describe the interaction between electromagnetic fields and particles.
- Naive approach: add up all particles using atomics.



Particle to Grid Reduction



Particle to Grid Reduction

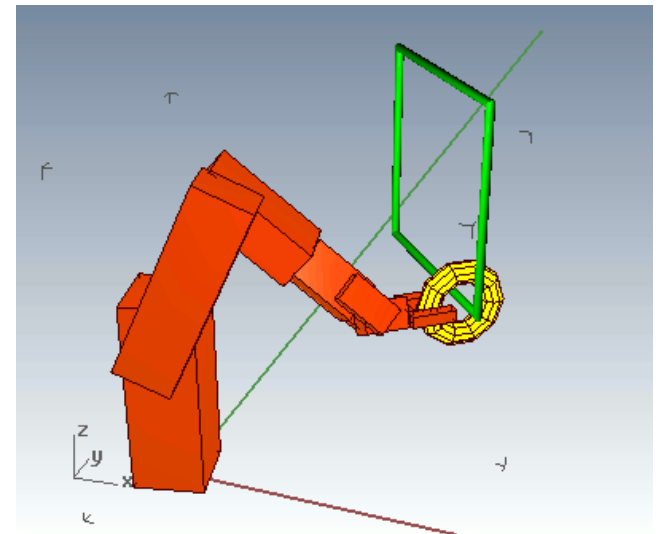
- Problem also encountered in gravitational N-Body simulations.
- Solution available, but they depend on GPU architecture → loss of portability.
- FPGAs perfect for this kind of task → limited availability.

Conclusion

- OpenCL offers many advantages in computational science.
 - deployment on heterogenous systems,
 - seperation between "physics" and (architecture dependent) host code.
- But it is difficult getting started:
 - lack of documentation/examples,
 - limited debug possibilites.

Systems of linear Equations

- Systems of linear equations are used in:
 - solving of implicit differential equations,
 - inverse kinematics,
 - computer vision (OpenCV).
- Common methods:
 - Gauß-Seidel-Methods (SOR,...),
 - Conjugate Gradient Method (CG),
 - Cholesky-Method.



System of linear Equations

