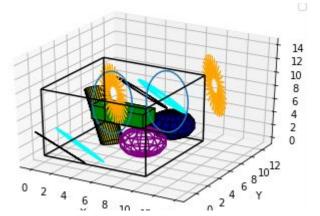
Accelerating the particle-in-cell method of plasma and particle beam simulation using CUDA tools.

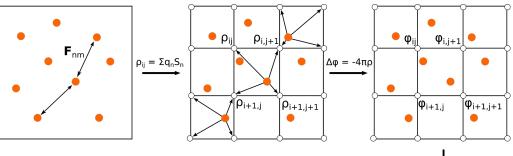
Ivan Kadochnikov - kadivas@jinr.ru

Ef an ef_python

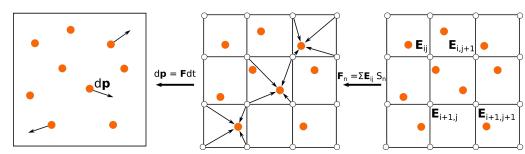
- Low-energy particle simulation using the Particle-in-Cell (PIC) method
- Open-source
- Focus on modeling ion sources and particle beams
- Support for external fields, conducting volumes and particle generators



Particle-in-cell



 $\mathbf{E} = -\nabla \phi$



Particle mover

• Leapfrog second-order explicit method (Boris scheme)

$$\begin{aligned} \mathbf{r}_{i+1} &= \mathbf{r}_i + \mathbf{v}_{i+1/2} \Delta t \\ \mathbf{v}_{i+1/2} &= \mathbf{v}_+ + \frac{q \Delta t}{2m} \mathbf{E}(\mathbf{r}_i) \\ \mathbf{v}_+ &= \mathbf{v}_- + \frac{q \Delta t}{2mc} (\mathbf{v}_+ + \mathbf{v}_-) \times \mathbf{B}(\mathbf{r}_i) \\ \mathbf{v}_- &= \mathbf{v}_{i-1/2} + \frac{q \Delta t}{2m} \mathbf{E}(\mathbf{r}_i) \end{aligned}$$

Field solver

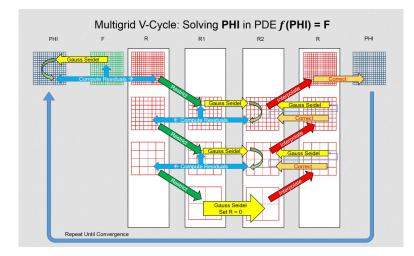
- Poisson equation (no dynamic magnetic field) $\Delta \phi = -4\pi \rho$
- Finite Difference Method (FDM) on a rectangular grid
- Initially solved by conjugate gradient method with scipy.sparse.linalg.cg

Each time step

- Push particles
 - Get electric field at particle positions(grid-to-particle, linear interpolation)
 - Push particle positions
- Generate new particles
- Solve field
 - Evaluate collective charge density (particle-to-grid)
 - Solve poisson equations for potential
 - Compute field from potential (just gradient)
- Prepare new particles
 - Get field at new particles(grid-to-particle, linear interpolation)
 - Set new particle velocities half a time step back

Algebraic Multi-Grid solver

• Multi-scale methods for faster FDM solving



PyAMG and AMGX





Non-GPU improvements

- Use diagonal sparse matrix functions of Scipy
- Caching inner nodes
- Grid-to-particle and particle-to-grid with numpy methods

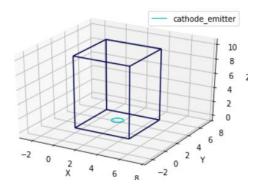
Numpy to Cupy

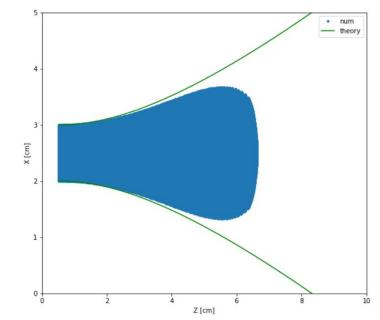
- Intended as a drop-in replacement/upgrade
- Some methods have new names
- No interpolator class
 - Custom CUDA kernel



Axially symmetric beam contour

- 50 x 50 x 100 grid
- 100 steps
- 5000 x 100 particles





Results

- 40x40x5000 grid, 2800 steps, 4x2800 particles
 - Simulated in 20 minutes on GPU (resources provided by JINR HybriLIT cluster)
- GPU accelerated every step of the simulation process
 - Optional, by default using pyamg and numpy
 - Except output and equation matrix creation
- 200+ tests with 91% coverage
- Next step:
 - OpenCL
 - MPI

Thank you!

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