

Scalable, Data Driven Plasma Simulations with

PICon GPU



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hzdr

Platform for Advanced Scientific Computing Conference (PASC19)

Zürich (Switzerland), *June 13th 2019*

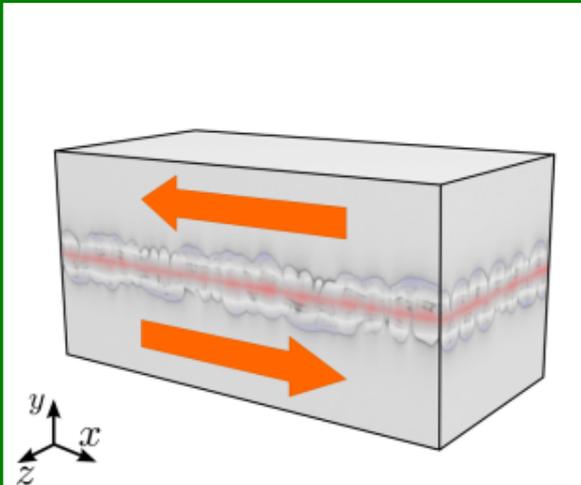


**HELMHOLTZ
ZENTRUM DRESDEN
ROSSENDORF**

Laser-Plasma Physics

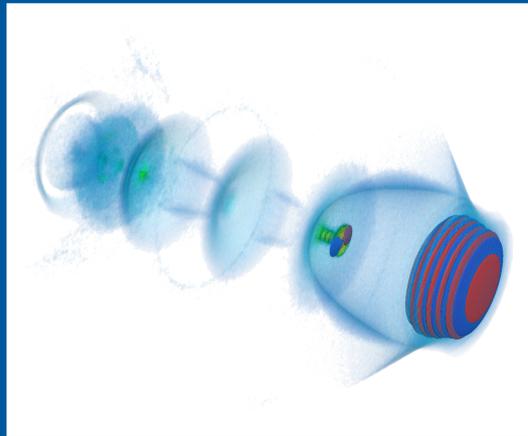
Plasma Instabilities

- Astrophysics
- Control of Laser-Plasmas



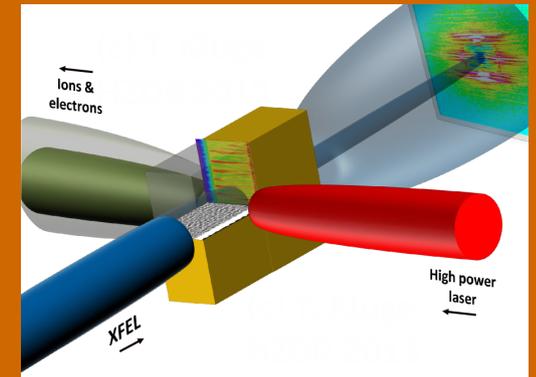
Electron Acceleration with Lasers

- Compact X-Ray sources
- Push the Energy Frontier



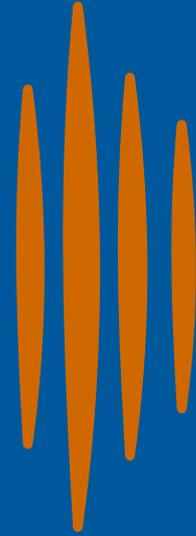
Ion Acceleration with Lasers

- Compact Ion Sources
- HED Physics

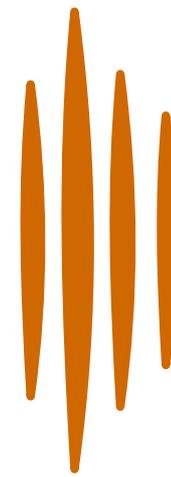


PICon GPU

Scalability



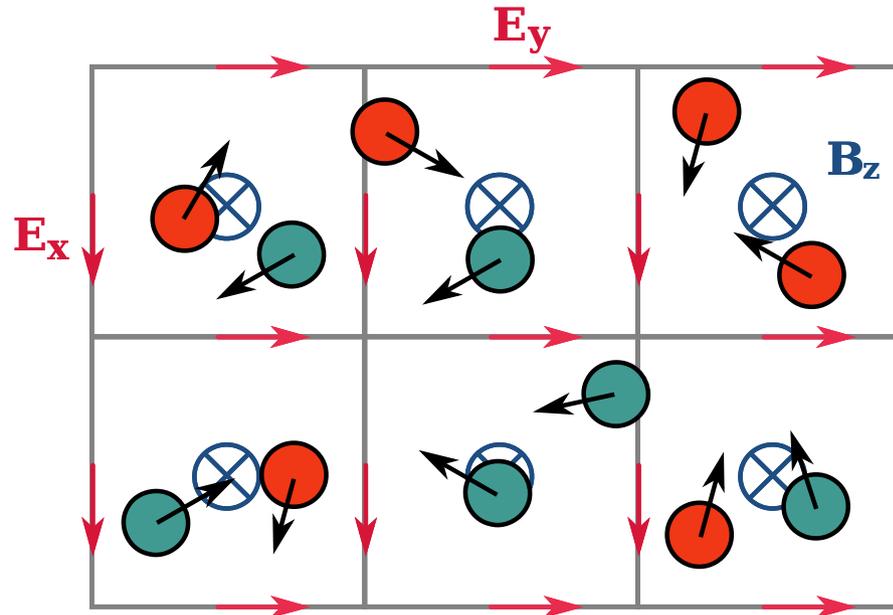
PICon GPU



- Eulerian: electro-magnetic fields
- Lagrangian: particles in Vlasov-equation

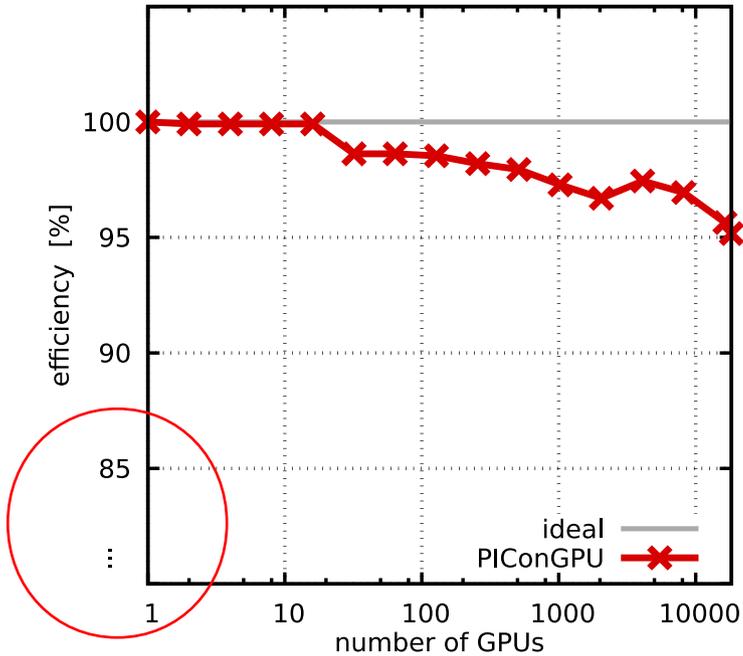
resolve:

- ω_{pe}
- ω_c

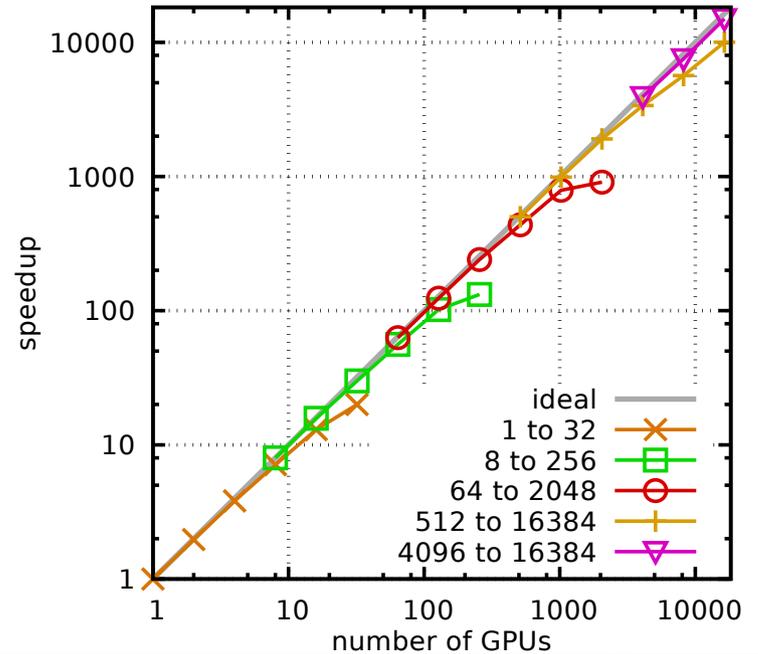


3D 3V

Weak Scaling



Strong Scaling

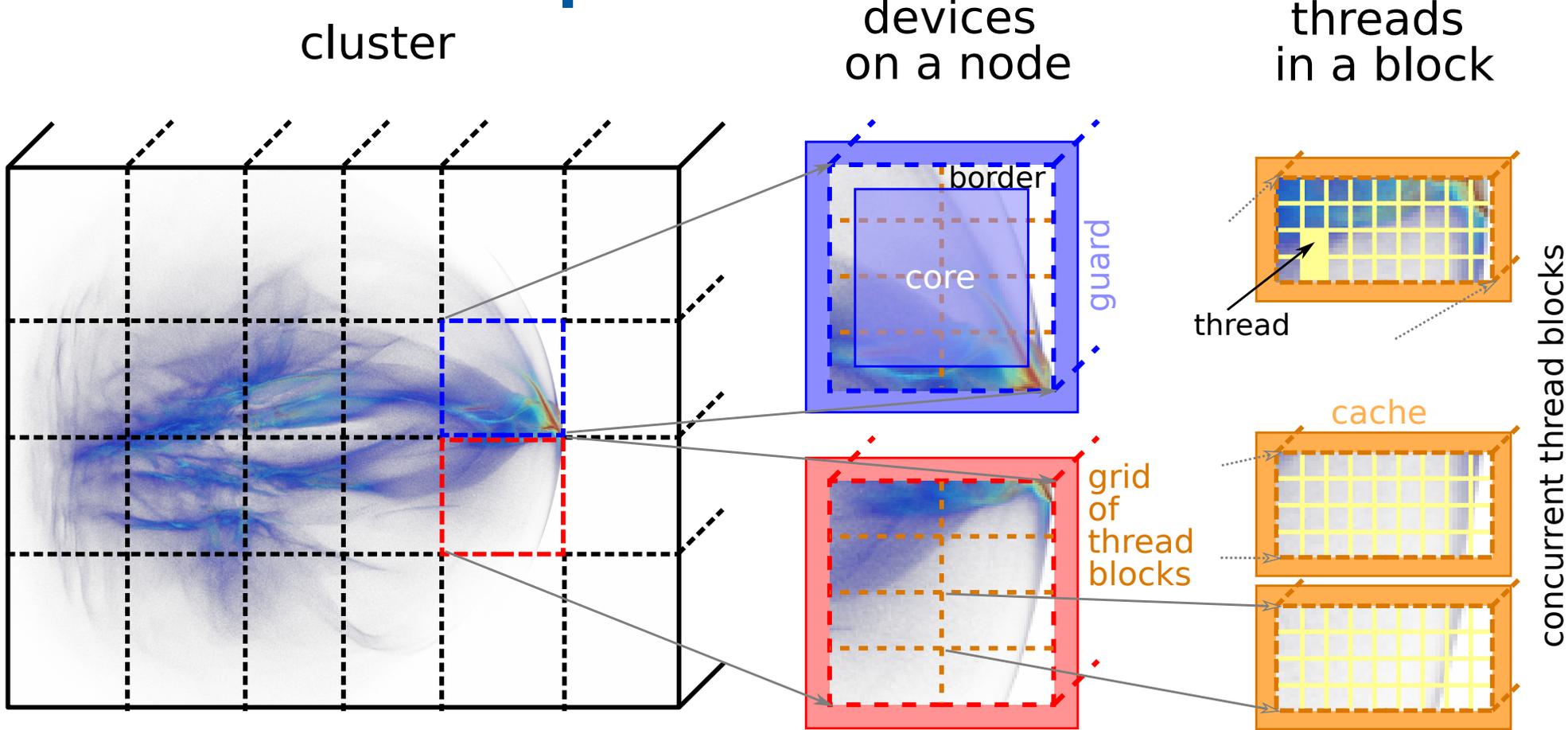


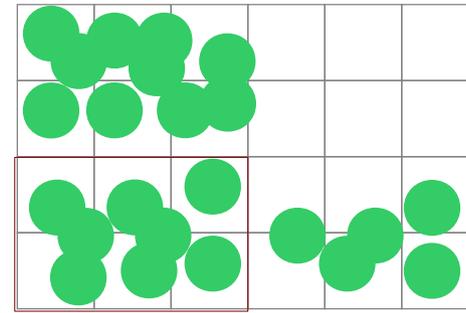
M. Bussmann, A. Huebl et al., SC'13,
DOI:10.1145/2503210.2504564



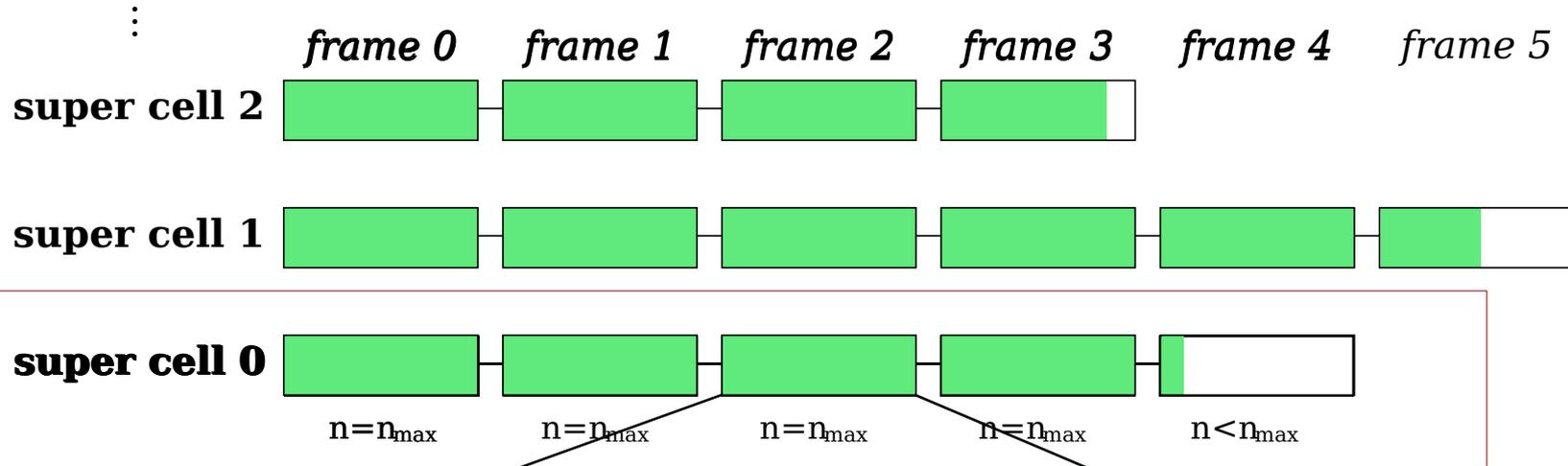
7.2 PFlop/s (DP) + 1.4 PFlop/s (SP)

Domain Decomposition

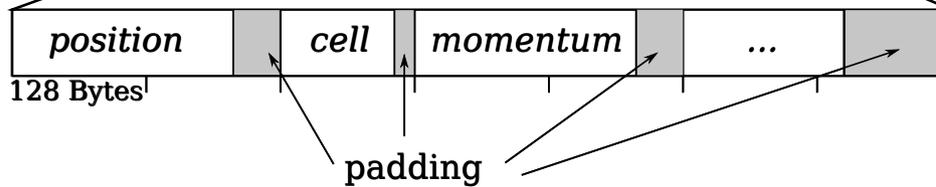




$$\vec{E}, \vec{B}$$



attributes of n particles



H. Bura et al., IEEE Trans. Plasma Sci. (2010), DOI:10.1109/TPS.2010.2064310

M. Bussmann et al., SC'13 (2013), DOI:10.1145/2503210.2504564

A. Huebl, Diplomarbeit (2014), DOI:10.5281/zenodo.15924

HPC Application Software Stack

Application

Containers and Algorithms

helper

In-Node Acceleration

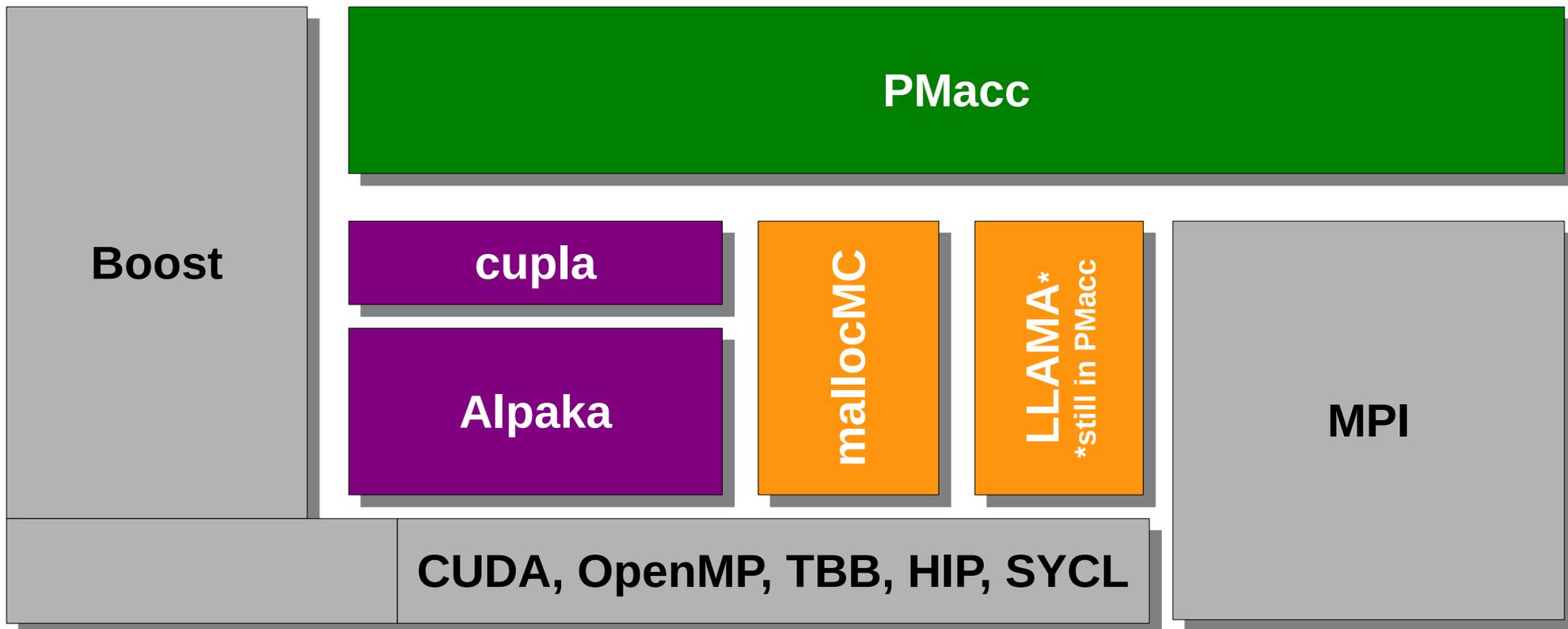
**Message-
Passing**

PICon GPU



Plugins

I/O coupling





alpaaka

Performance Portability

Parallel Programming Models

- re-implementation (Nx)
 - not affordable for domain-science teams

```
#ifdef CUDA_ENABLE
// CUDA Kernel implementation
// ...

#elif OPENMP_ENABLE
// OpenMP implementation
// ...

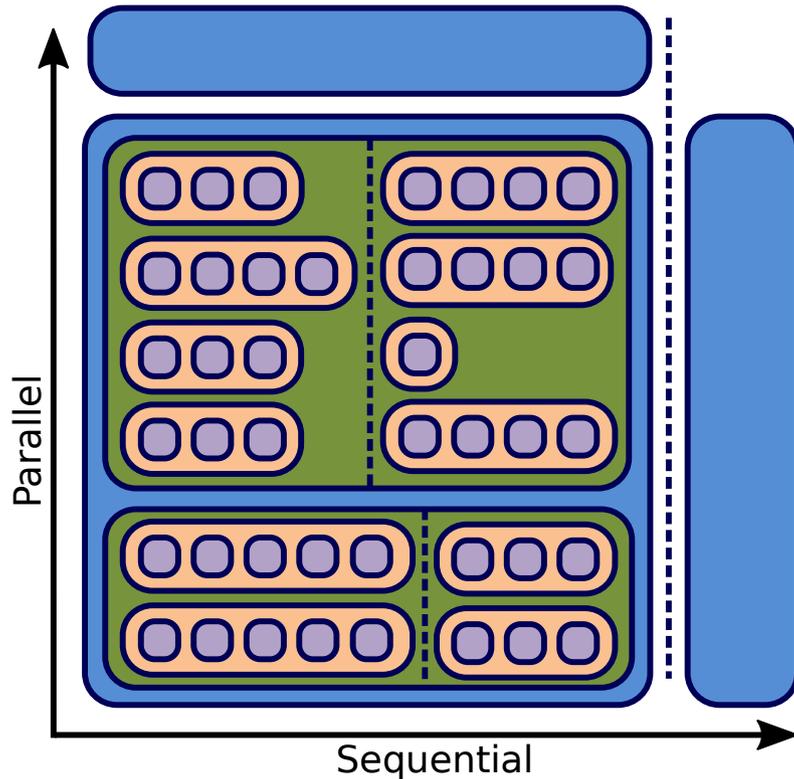
#else
// Sequential CPU implementation
// ...

#endif
```

- single-source manycore
 - CUDA: CUDAx86 (PGI), GPUOcelot (GaTech); today: HIP (AMD)
 - *Abstract & configure the task parallelism!*



Alpaka's Task Parallelism



whole parallel collection



fully independent part
of the grid



executed concurrently

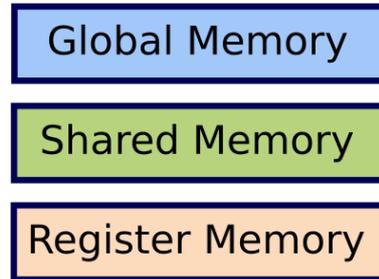
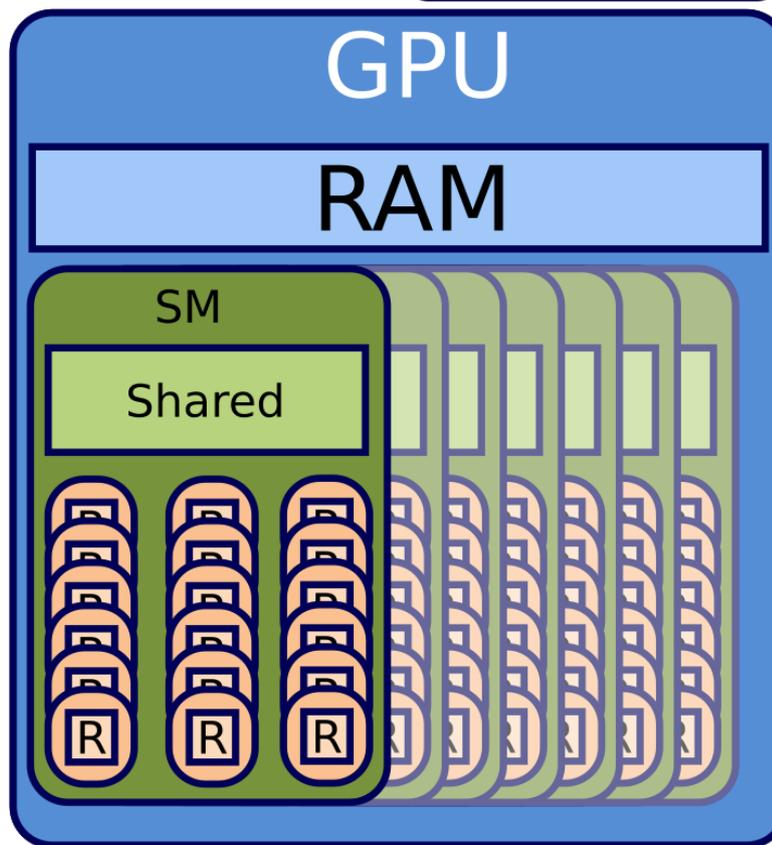
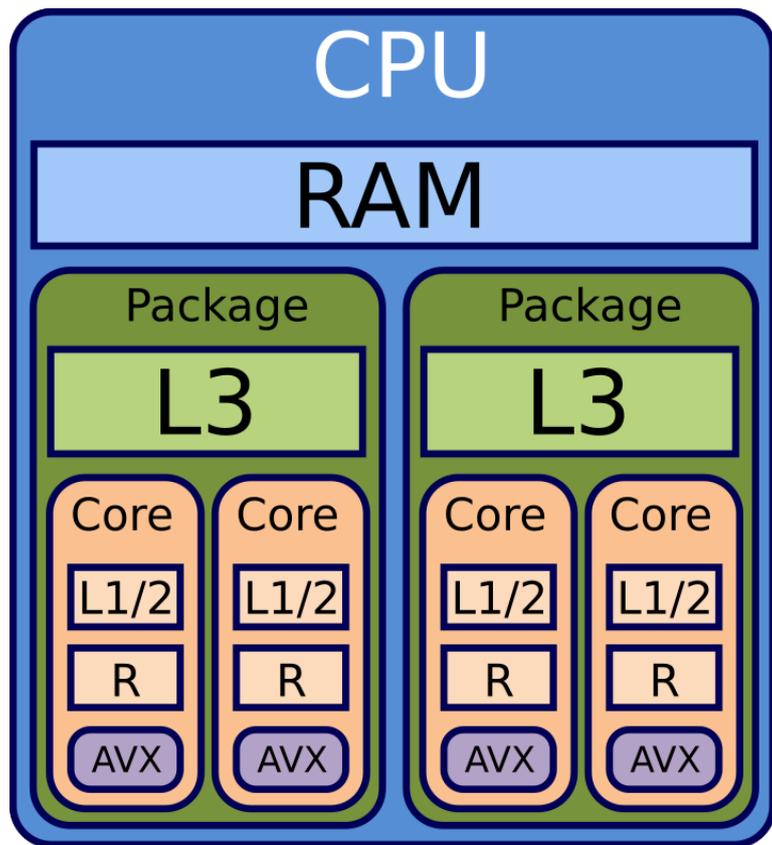


sub-thread, sequential
lock-step

---- Synchronize

Alpaka

Hardware Mapping



OpenMP2

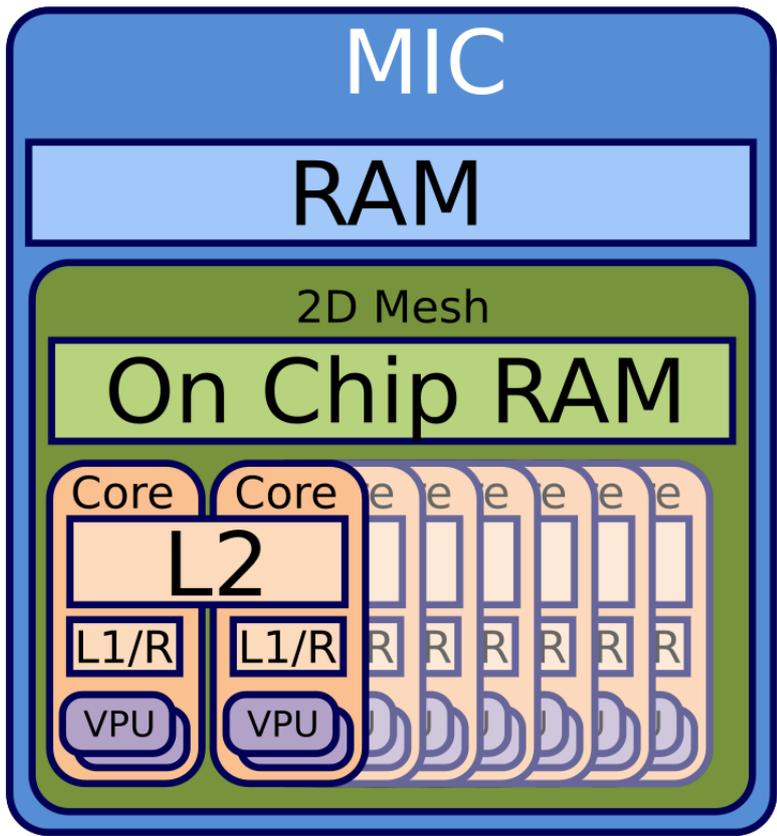
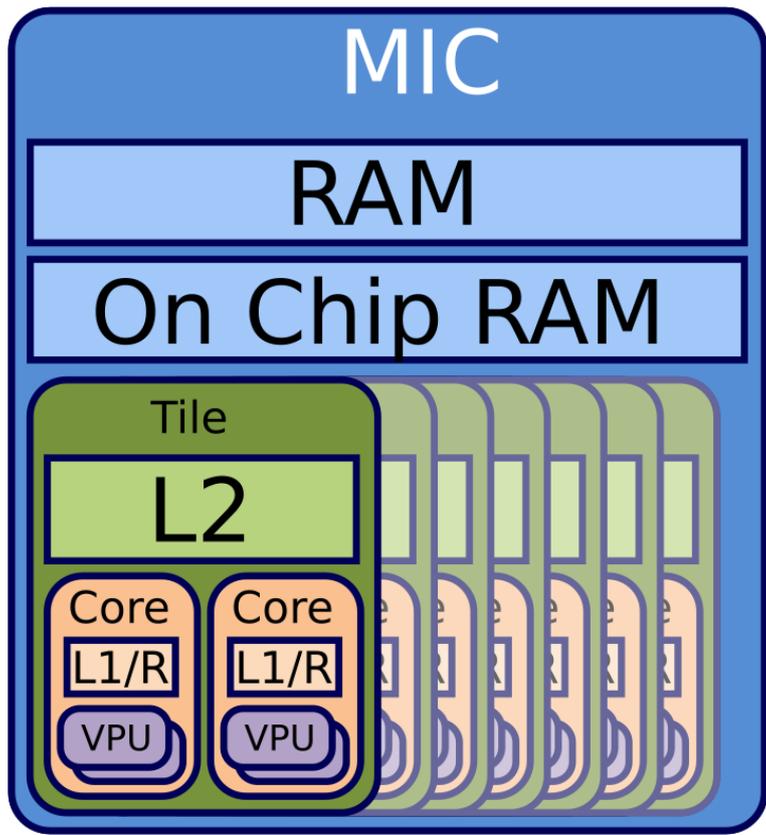
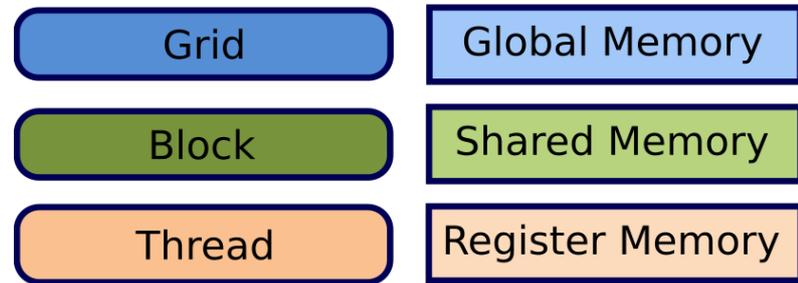
Threads

CUDA

HIP

Alpaka

... e.g. MIC



OpenMP2

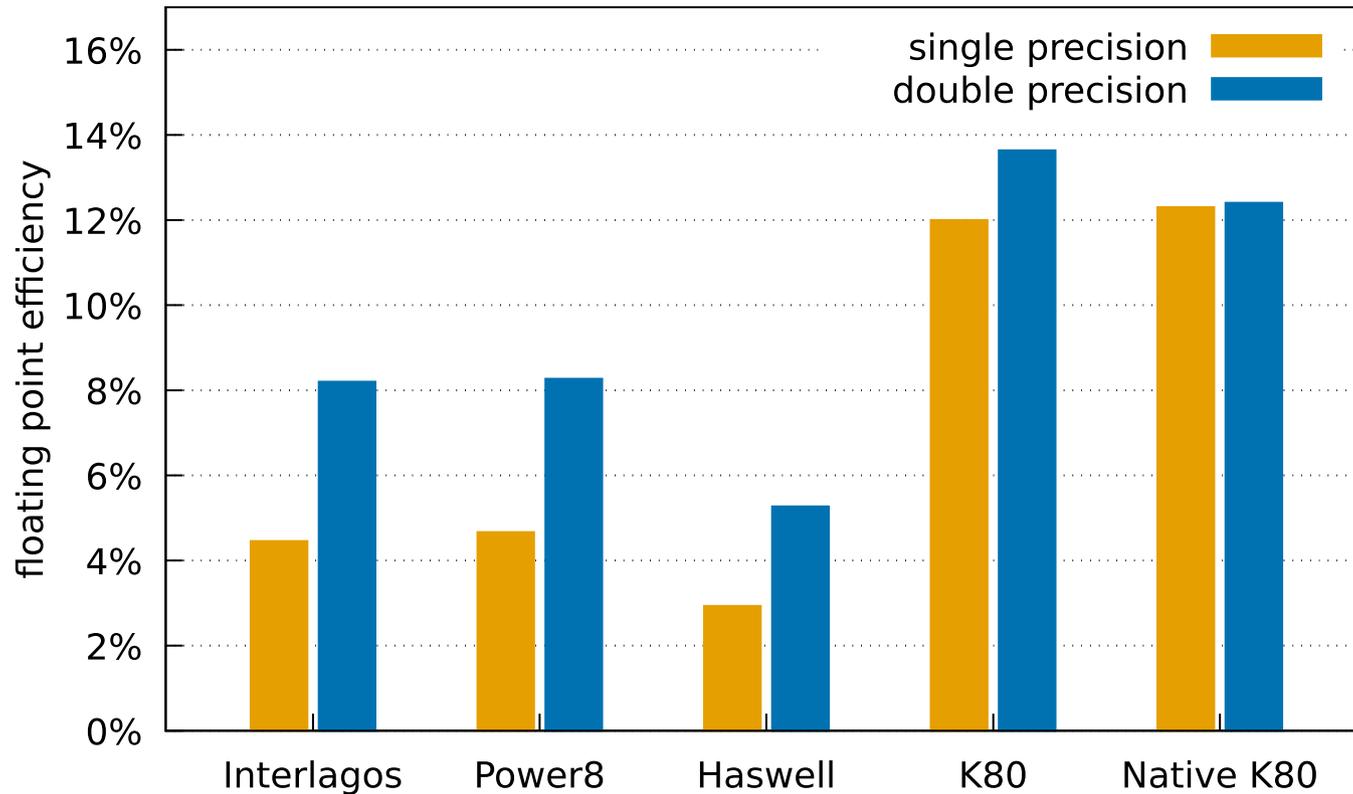
TBB

Threads

SYCL

PIConGPU on Alpaka

Performance after only 3 (!) weeks



Maintainability
PIConGPU +
PMacc code lines

Before: 80k LOC
(20k in kernels)
After: 50k LOC
(1 year)

E. Zenker et al., ISC (2016), DOI:10.1007/978-3-319-46079-6_21

A. Matthes et al., ISC (2017), DOI:10.1007/978-3-319-67630-2_36



Scientific Data Workflows

The actual HPC challenge

The I/O Challenge Continues!

www.olcf.ornl.gov/frontier/

SYSTEM SPECS	TITAN	SUMMIT	FRONTIER
Peak Performance	27 PF	200 PF	> 1.5 EF
Storage	32 PB, 1 TB/s, Lustre Filesystem	250 PB, 2.5 TB/s, GPFS™	2-4x performance and capacity of Summit's I/O subsystem. Frontier will have near node storage like Summit.

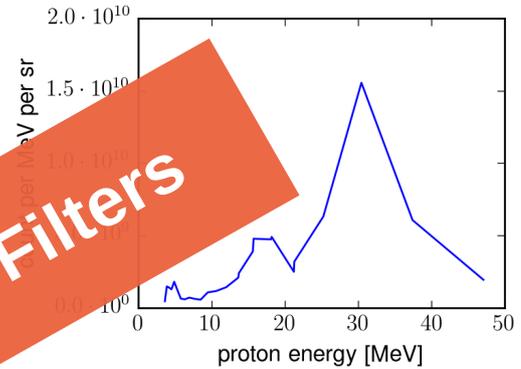
1/3x

1/3x

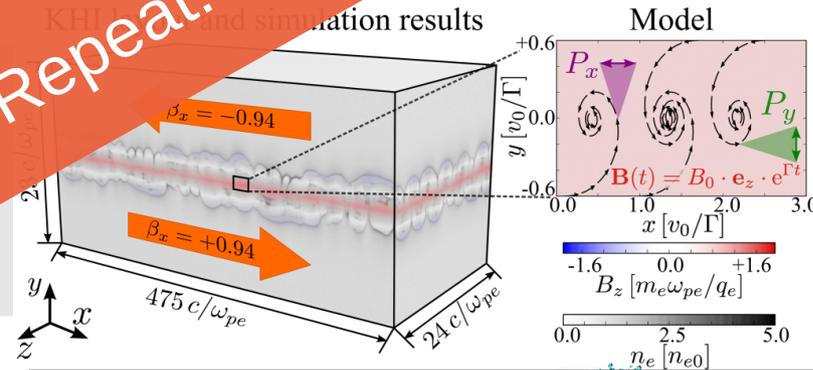
In situ approaches: **tightly** versus **loosely** coupled workflows

Virtual Detectors: Tightly Coupled Plugins

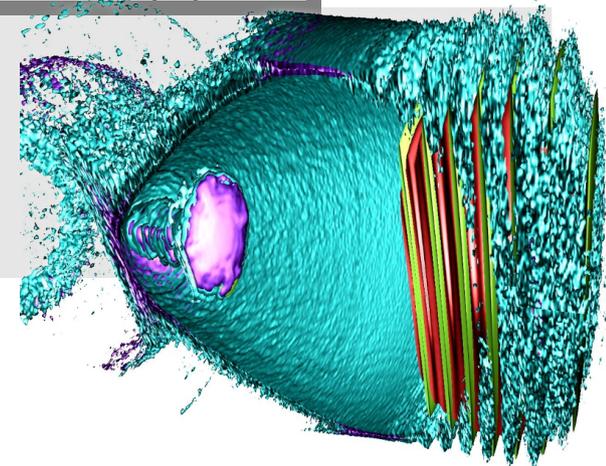
Binning of a spectrogram
Creation of a phase space image



In situ radiation diagnostics



Ray-cast or photo-realistic ray-trace
Lossy data compression

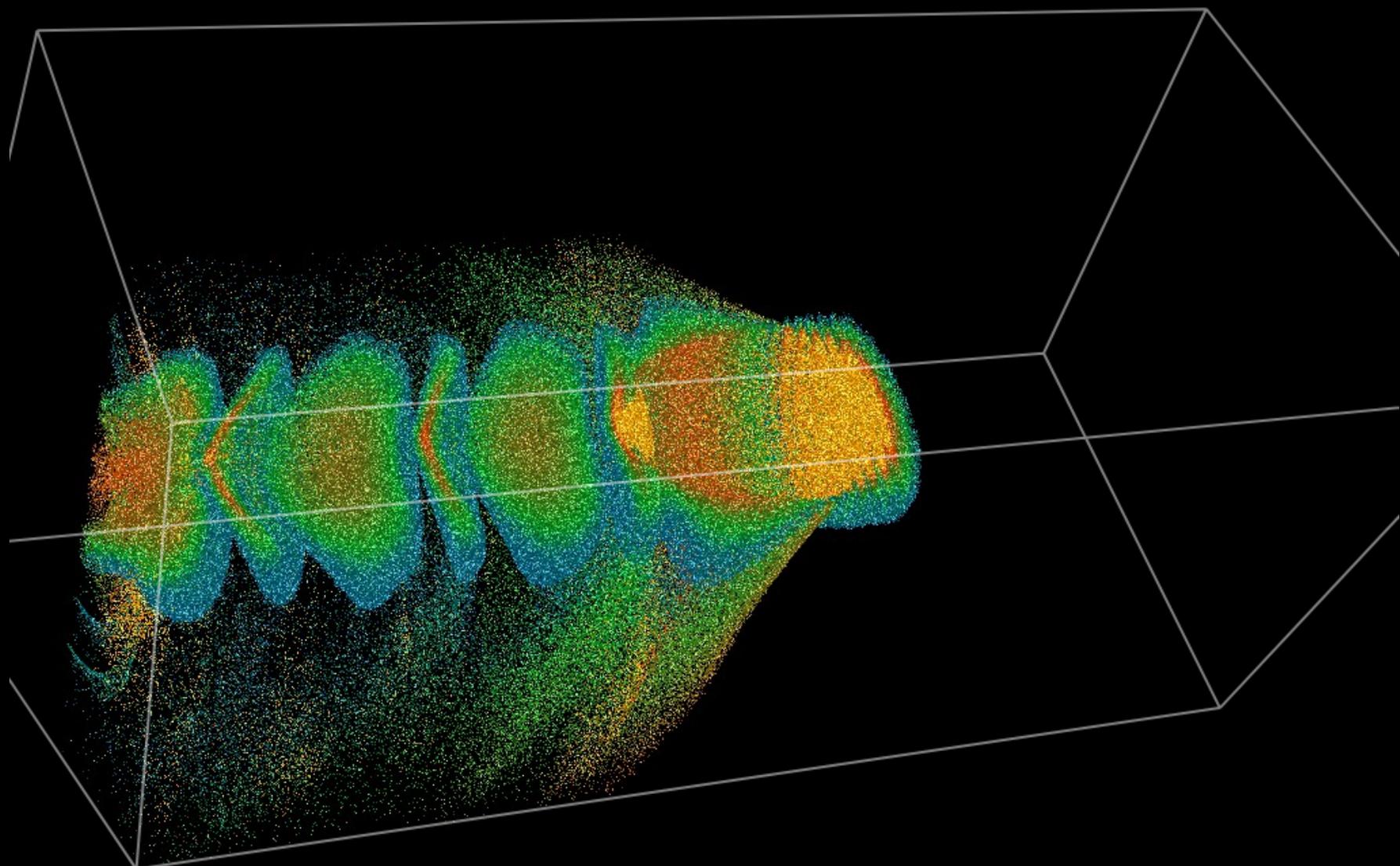


Observe, Correlate & Repeat: Particle Filters

- A. Huebl et al. (2014), DOI:10.1109/TPS.2014.2327392
- R. Pausch et al. (2017), DOI:10.1103/PhysRevE.96.013316
- A. Matthes, A. Huebl et al., ISC'16 (2016), DOI:10.14529/jsfi160403
- A. Huebl et al., ISC'17 (2017), DOI:10.1007/978-3-319-67630-2_2

0 ms Simulation
10 ms Rendering

118 Mio. Particles
59 Mio. Cells

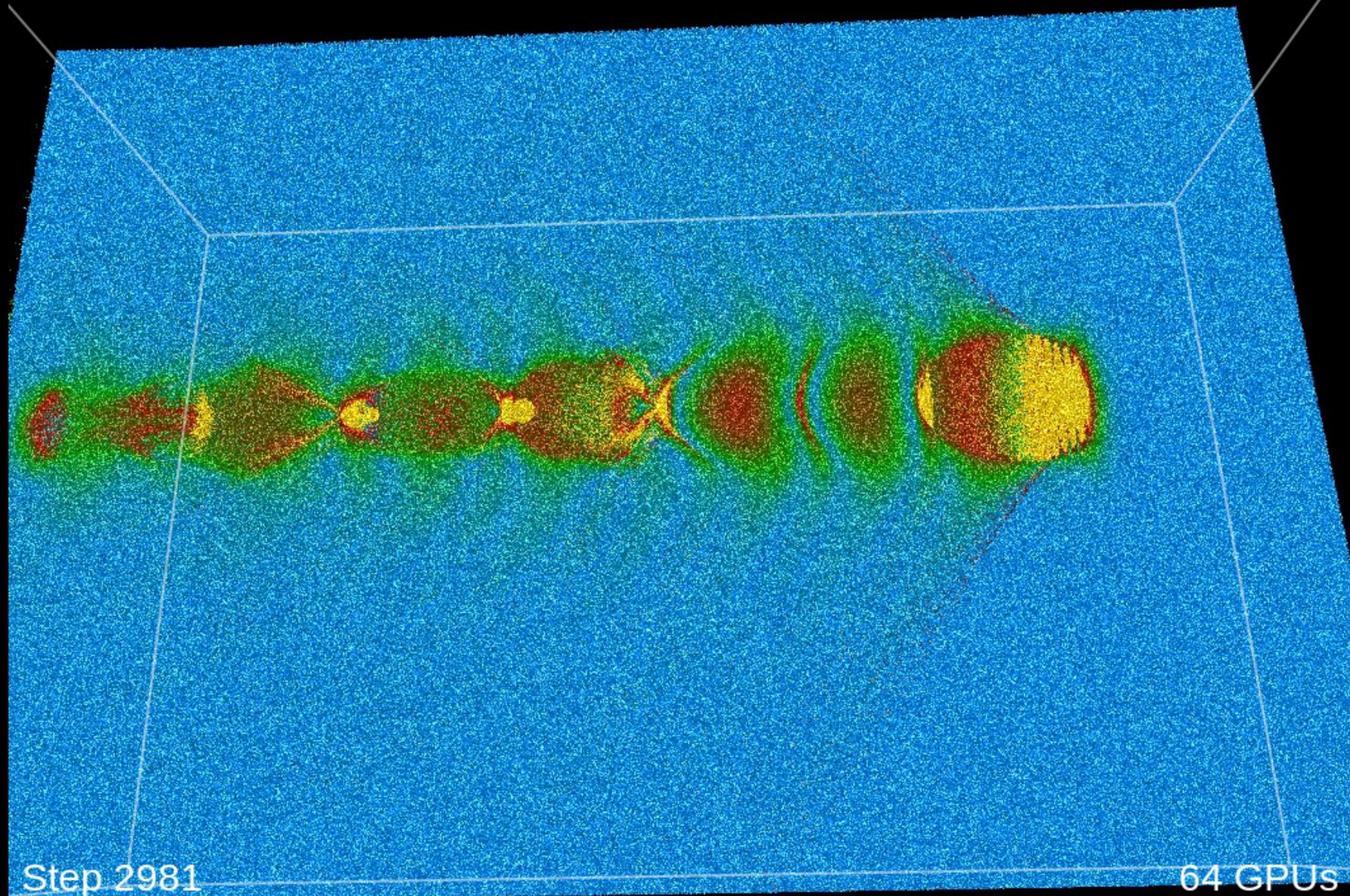


Step 3304

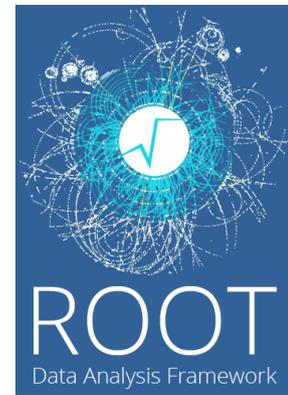
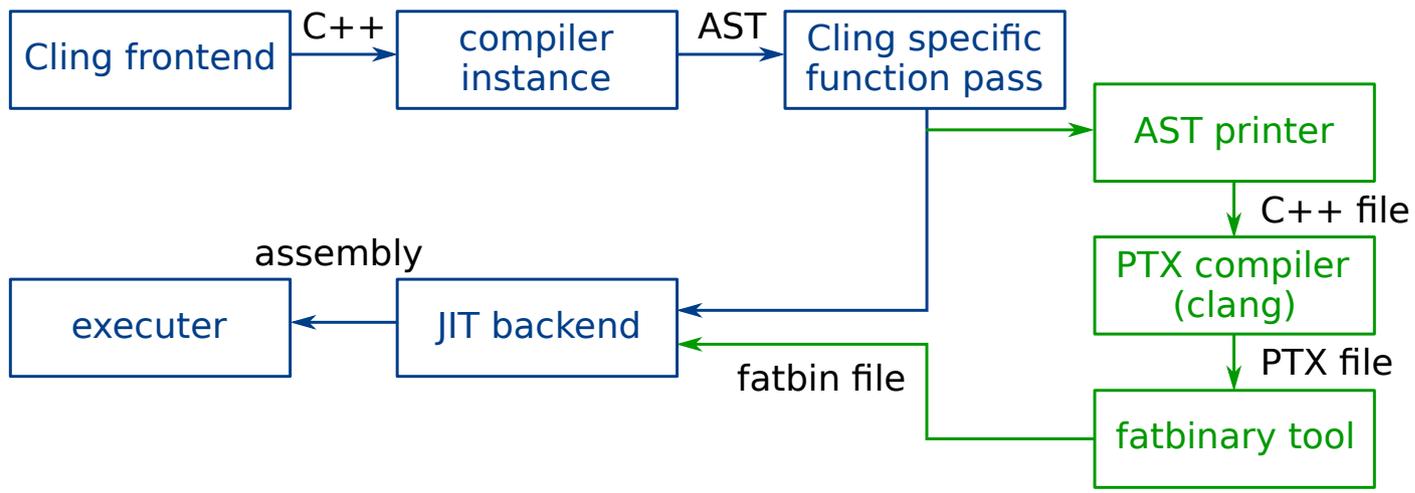
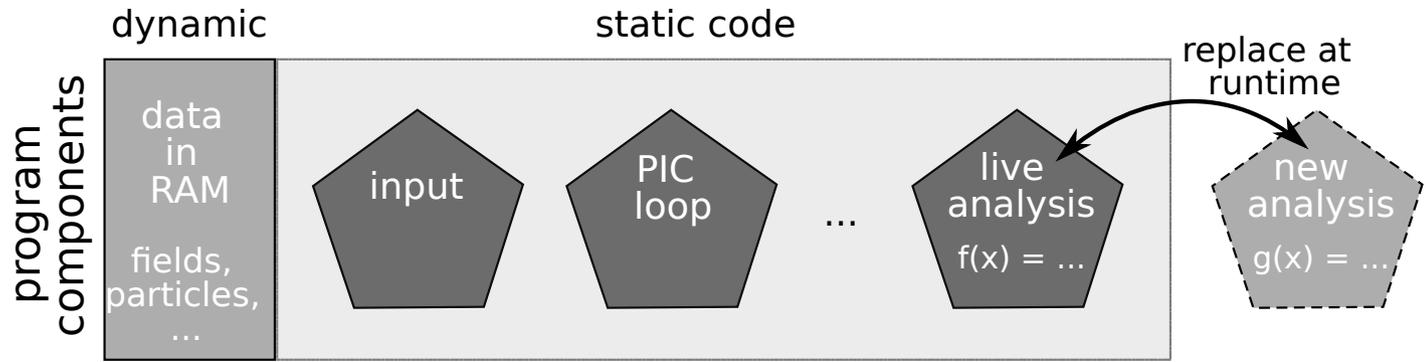
32 GPUs

243 ms Simulation
152 ms Rendering

1678 Mio. Particles
839 Mio. Cells



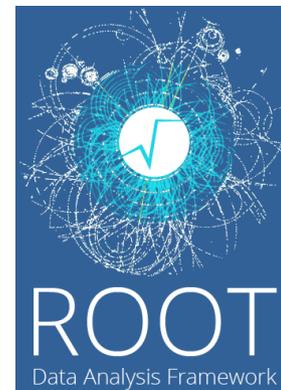
Cling CUDA



S. Ehrig (2018), HZDR/TU Dresden, Diploma Thesis



Cling CUDA: We implemented a GPU JIT!



jupyter CUDA_copy (autosaved)

Logout

File Edit View Insert Cell Kernel Widgets Help Trusted xeus-C++14-cuda



```
In [ ]: template <typename T>
__global__ void copy_kernel(T * in, T * out, unsigned int N){
    int id = blockIdx.x * blockDim.x + threadIdx.x;
    if(id < N)
        out[id] = in[in];
}
```

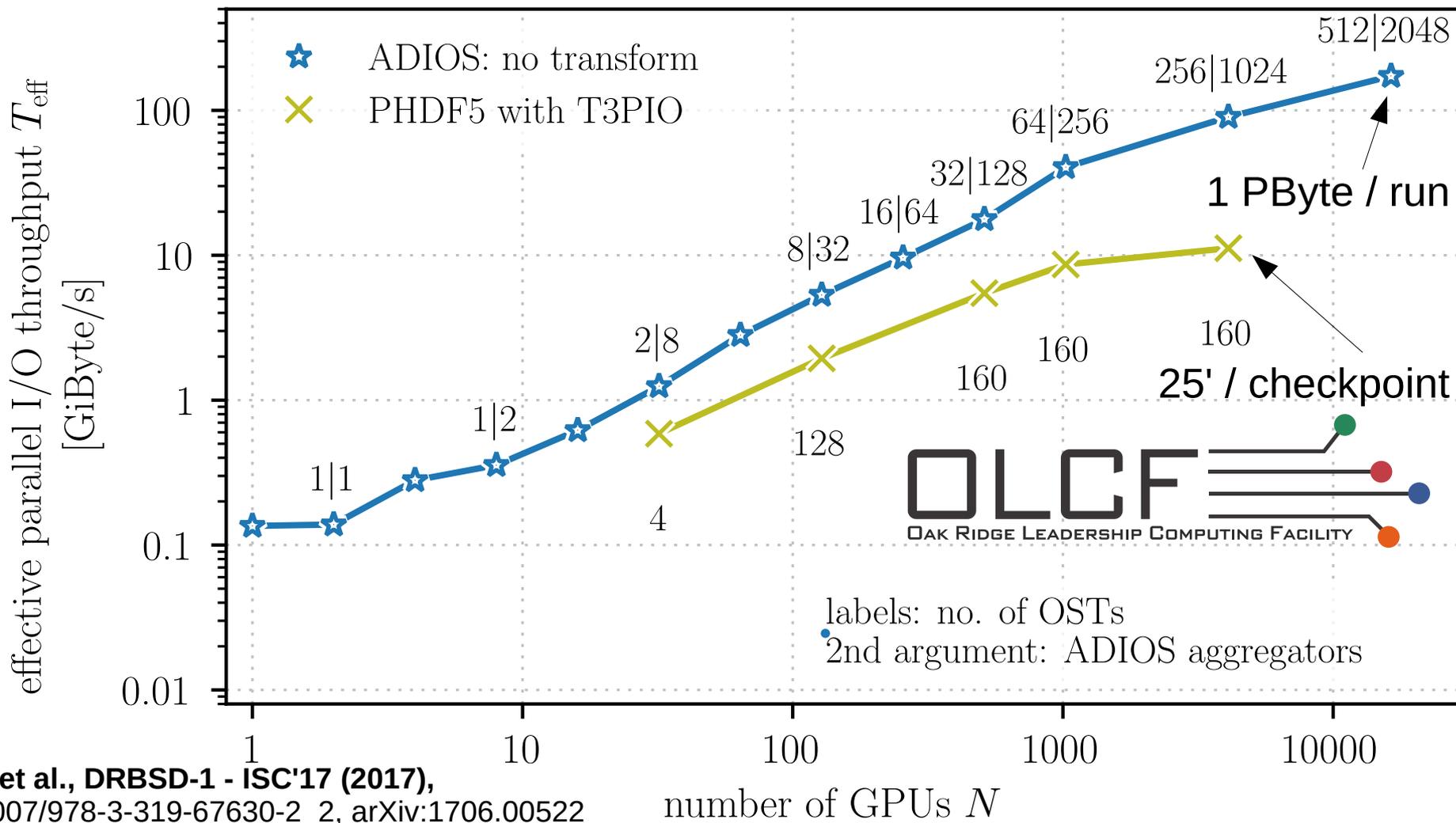


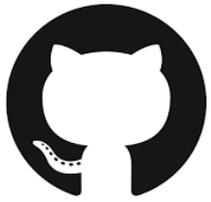
S. Ehrig (2018), HZDR/TU Dresden, Diploma Thesis

Exascale Challenge: I/O Scalability

Titan I/O Weak Scaling with PIconGPU

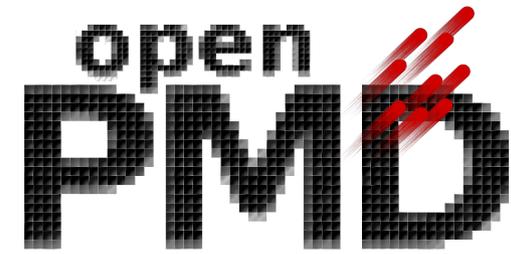
$$T_{\text{eff}} \equiv \frac{N \times S}{t_{\text{I/O}}}$$



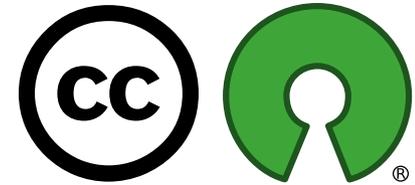


github.com/openPMD

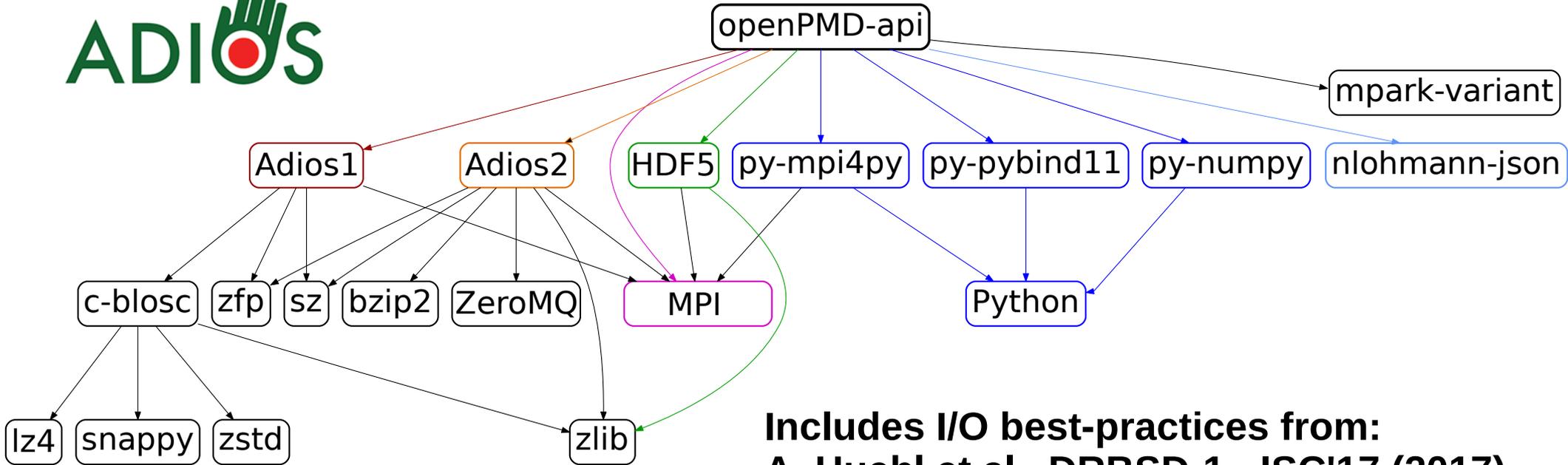
www.openPMD.org



- markup / schema for arbitrary hierarchical data formats
- truly, *scientifically* self-describing
- basis for open data workflows



Batteries included: openPMD-api



Includes I/O best-practices from:
A. Huebl et al., DRBSD-1 - ISC'17 (2017),
DOI:10.1007/978-3-319-67630-2_2, arXiv:1706.00522

Available via:



Spack



PyPI



Productivity

PIC as a Service

PIConGPU as a Service

- full focus on a science case: expressive, reproducible
 - *reduce* options to a *virtual experiment*
 - implement a **sharable data workflow** from best-practices
 - *transparently hide the HPC workflow*
- **collaborative service**: usability & productivity
 - accessible for **experimentalists** and **theoreticians**
 - **hackable** for power-users
 - **auto-documenting** and **auto-archiving**
 - ... **ML, digital twins, etc.**

Authors: S. Rudat, S. Starke, J. Kelling, A. Huebl
A. Huebl, G. Juckeland et al., SIAM'18 (2018)



RO DARE
ROSSENDORF DATA REPOSITORY



PICongPU as a Service Demo

Combine Qualitative and Quantitative Analysis



https://www.hzdr.de/db/Cms7p0... 240%

HZDR Internal

PICongPU

This site provisions a Jupyter notebook to configure and interact with PICongPU runs. Press the button to start your instance. It will be accessible through an open port on hypnos5.

Start ClusterJob Form

Start Jupyter Notebook

jupyter interactive (autosaved)

File Edit View Insert Cell Kernel Widgets Help Trusted Python 3

Scan: scan_0009 Refresh Sim Status Abort scan_0009

Simulation: sim_0001 (running)

resolution laser target

BASE_DENSITY_SI

Base_Density_SI [1/m^3] 23.00 - 23.00 + -

Visualize Download sim

Iteration 4300

Figure 1

section currently being reworked

Count

Energy

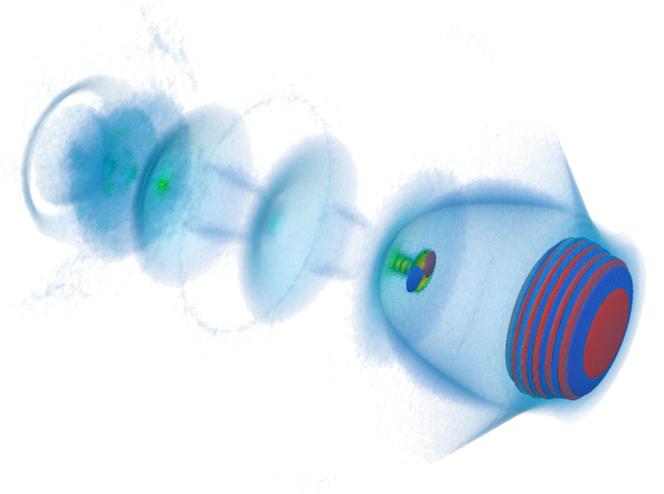
n (pH)

charge [Coulomb^-1]

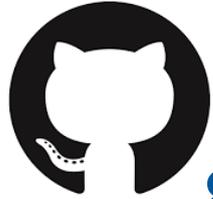
live

Summary

- **Scalability of Algorithms first**
 - Use abstract programming models
 - Use and build community libraries
- **In situ data reduction for all production jobs unavoidable**
 - Start loosely coupled, implement hot-spots tightly
 - Join the openPMD community for data description :)
- **Build intuitive, reproducible workflows**
 - Modular and transparent for power-users



talk by A. Huebl (HZDR) et al., a.huebl@hzdr.de



picongpu.hzdr.de



github.com/ComputationalRadiationPhysics

This project has been enabled by many people in open-source and open-science communities. Great thanks to the communities and developers of: PIconGPU, Alpaka, ROOT/Cling, Jupyter, the SciPy ecosystem, ADIOS, HDF5, Boost, CMake, openPMD, Spack, ...

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