

Extreme Throughput Data Taking, Management & Analysis

Real Life Experiences of Generating/Evaluating
Extreme Data Sets Around the World

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<https://www.fairfaxcounty.gov/news2/wp-content/uploads/2016/03/Fire-hydrant-with-water-blur.jpg>



<https://pluspng.com/img-png/bathtub-png-bathtub-png-1662.png>

Matter meets Information, DESY, Jan 14-15, 2019



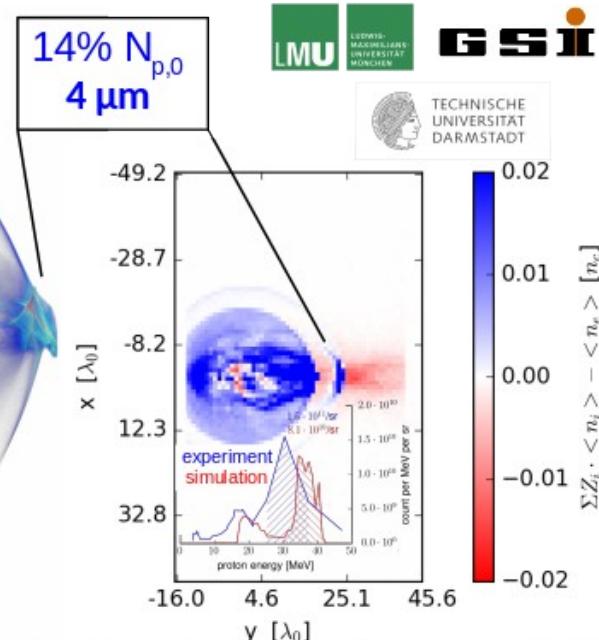
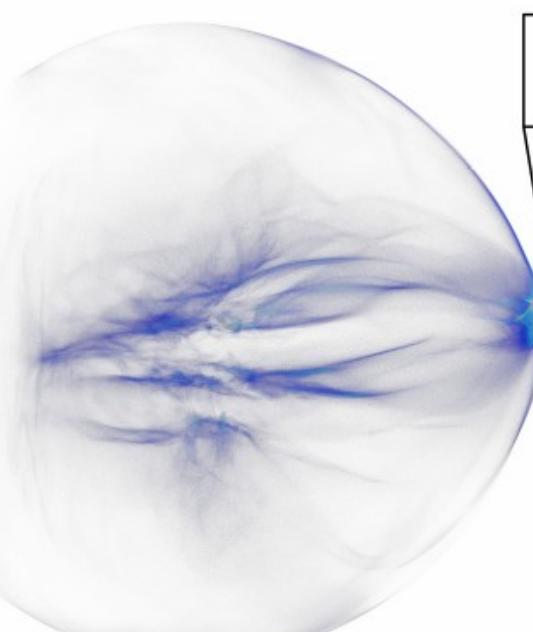
HZDR

 HELMHOLTZ
ZENTRUM DRESDEN
ROSSENDORF

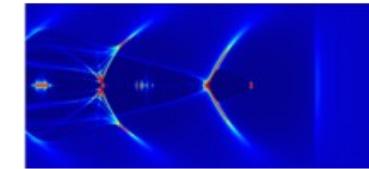
Our Background

well, one of many...

Laser Plasma Acceleration



e-: 4.2 GeV
in 9 cm



BNL (2014)

W. P. Leemans et al.,
PRL 113, 245002

P. Hilz, T. M. Ostermayr, A. Huebl et al., Nature Comm. 9, 423 (2018)

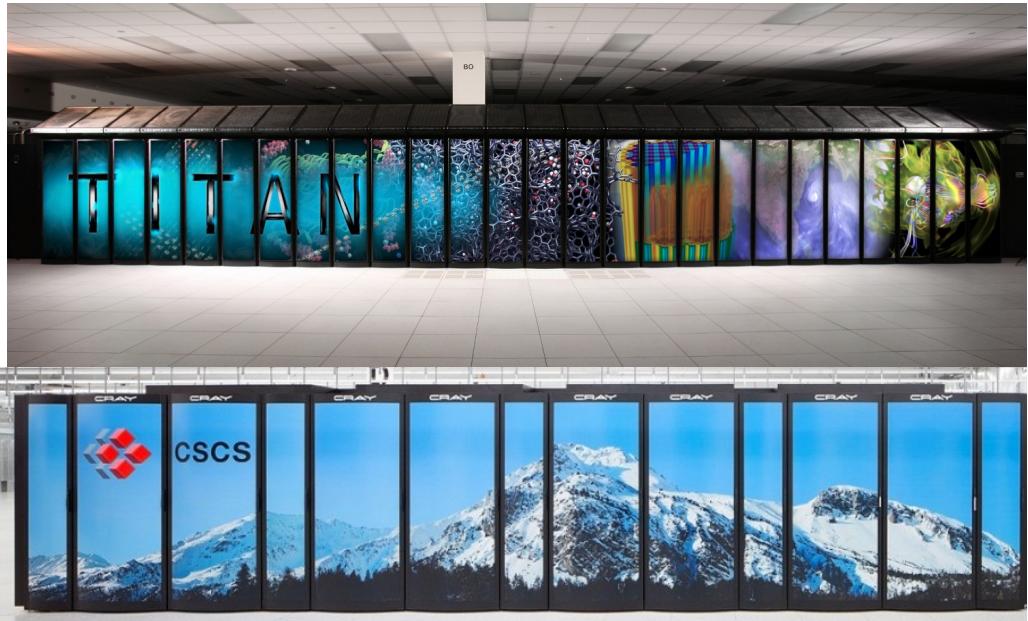


hZDR

Mitglied der Helmholtz-Gemeinschaft

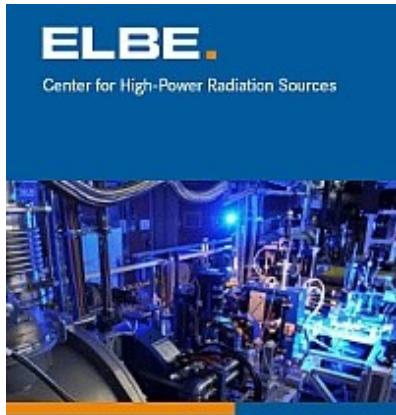
Particle-in-Cell Simulations a.k.a. digital twins

- Ab initio, electro-magnetic plasmas
- Scaling to the full-size of Titan & Piz Daint
- Gordon Bell finalist 2013



Mitglied der Helmholtz-Gemeinschaft

LK-II Facilities for Plasma Physics, Material Science,...



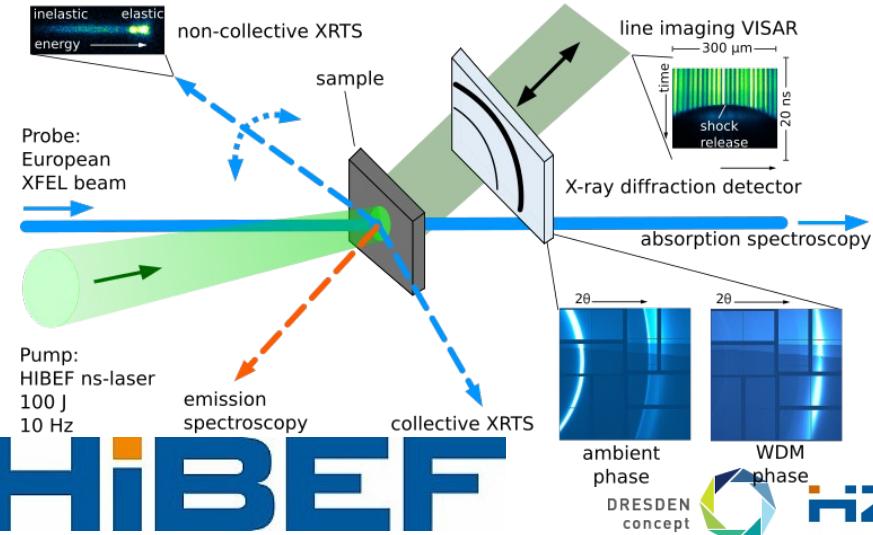
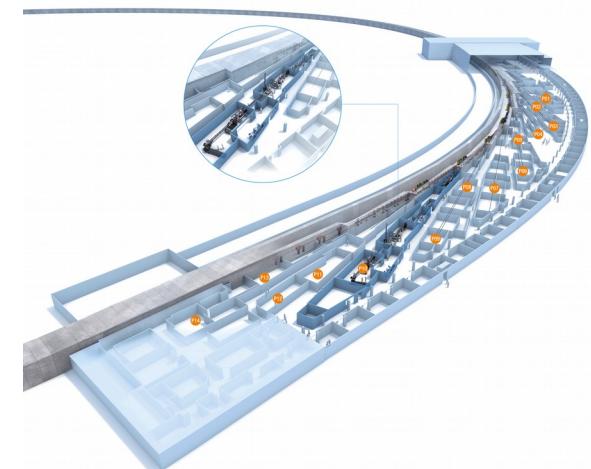
„USS Enterprise“ Experiments

- Novel methods, setups, (commodity) detectors
- Unprecedented bandwidth/latency requirements
- Photon Science different from HEP



HiBEF

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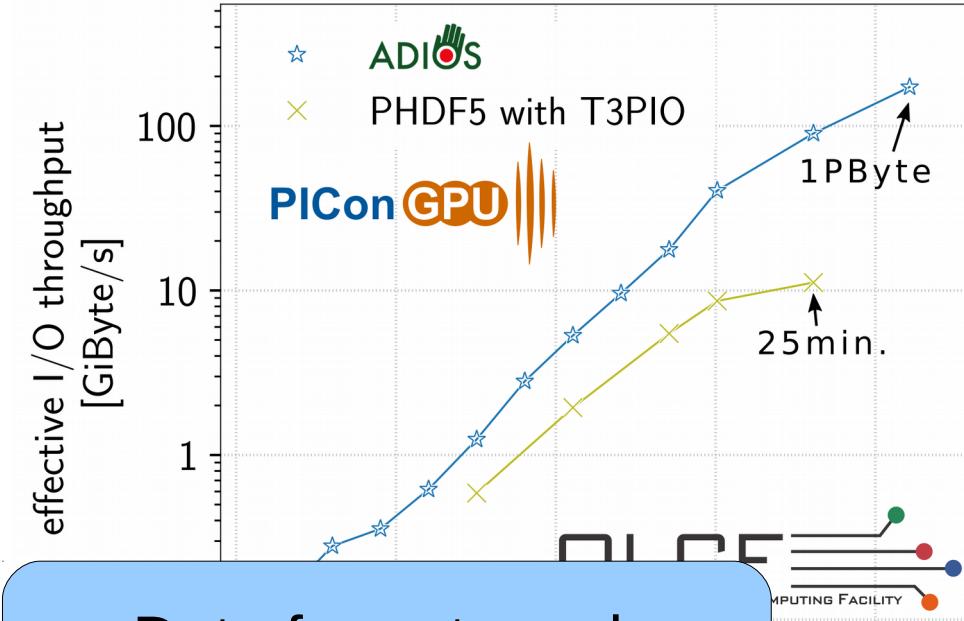
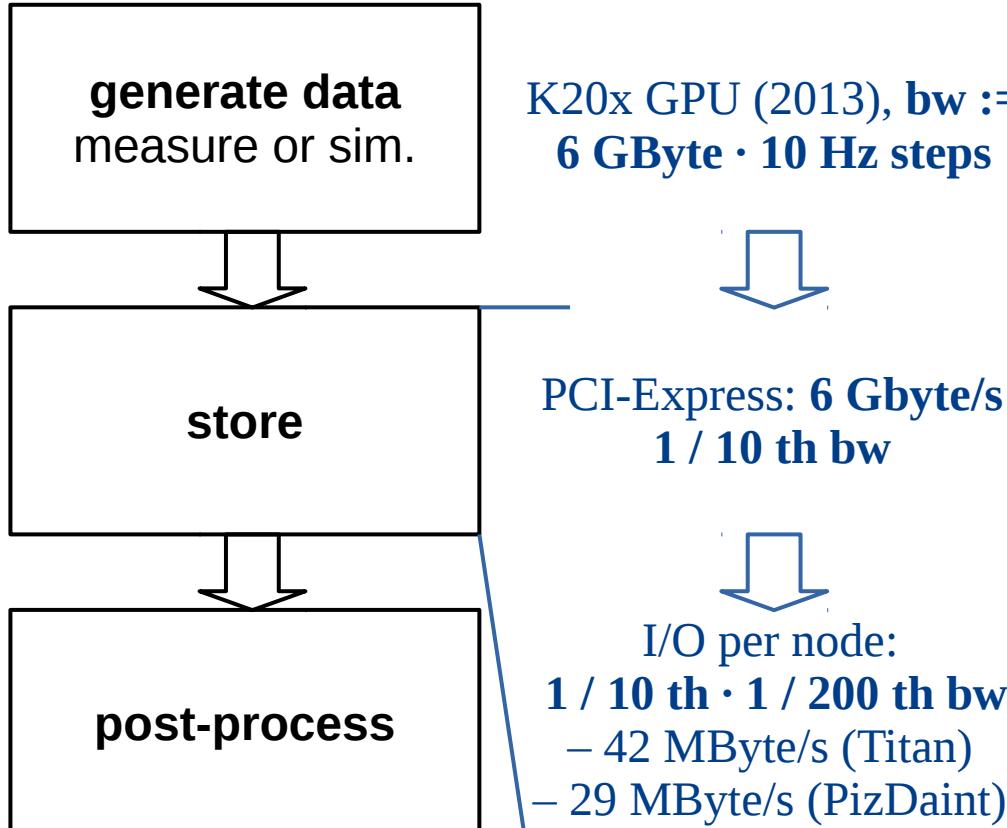


Data Challenges at Extreme Scale

(also coming to your site! Soon.)

Typical Post-Processing

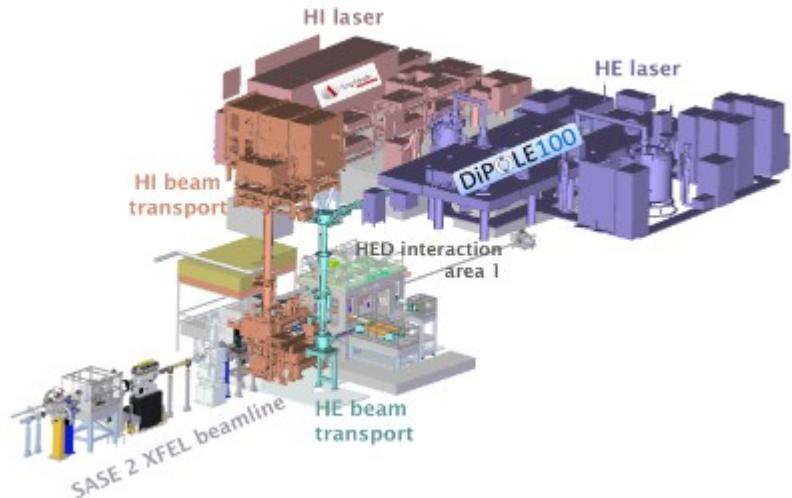
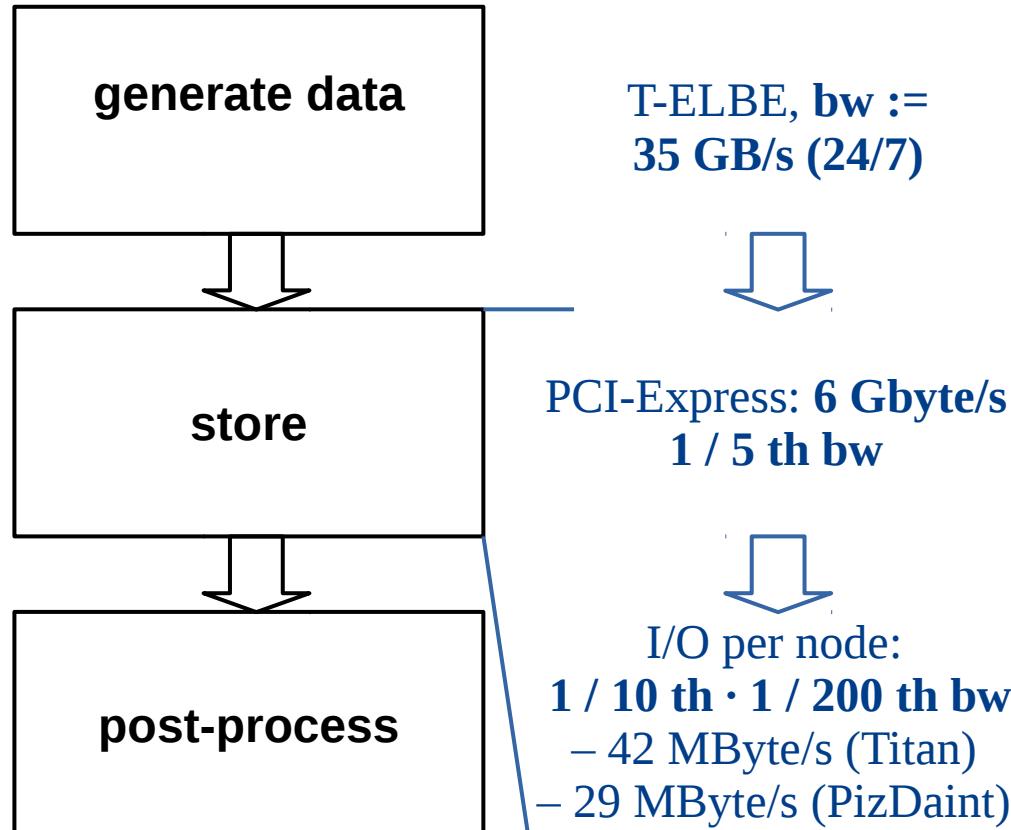
A. Huebl et al., DRBSD-1 - ISC'17 (2017),
DOI:10.1007/978-3-319-67630-2_2, arXiv:1706.00522



Data formats and
I/O libraries matter

Summit (ORNL, 2018): ratio 4x “worse” - gap of 10^4

Next-Generation Experiments



When the Bath-Tub is Full... - Site Storage Policies for >1 PB data sets

	Site A	Site B	Site C
Capacity	250 PByte	6 PByte	3 PByte
Capacity/FLOP	2 Byte/FLOP/s	0.3 Byte/FLOP/s	10 Byte/FLOP/s
Bandwidth	2.5 TB/s	100 GB/s (estd.)	40 GB/s
Bandwidth/FLOP	20 µByte/FLOP	5 µByte/FLOP	133 µByte/FLOP
Retention Time	90 days + archive	30 days	∞

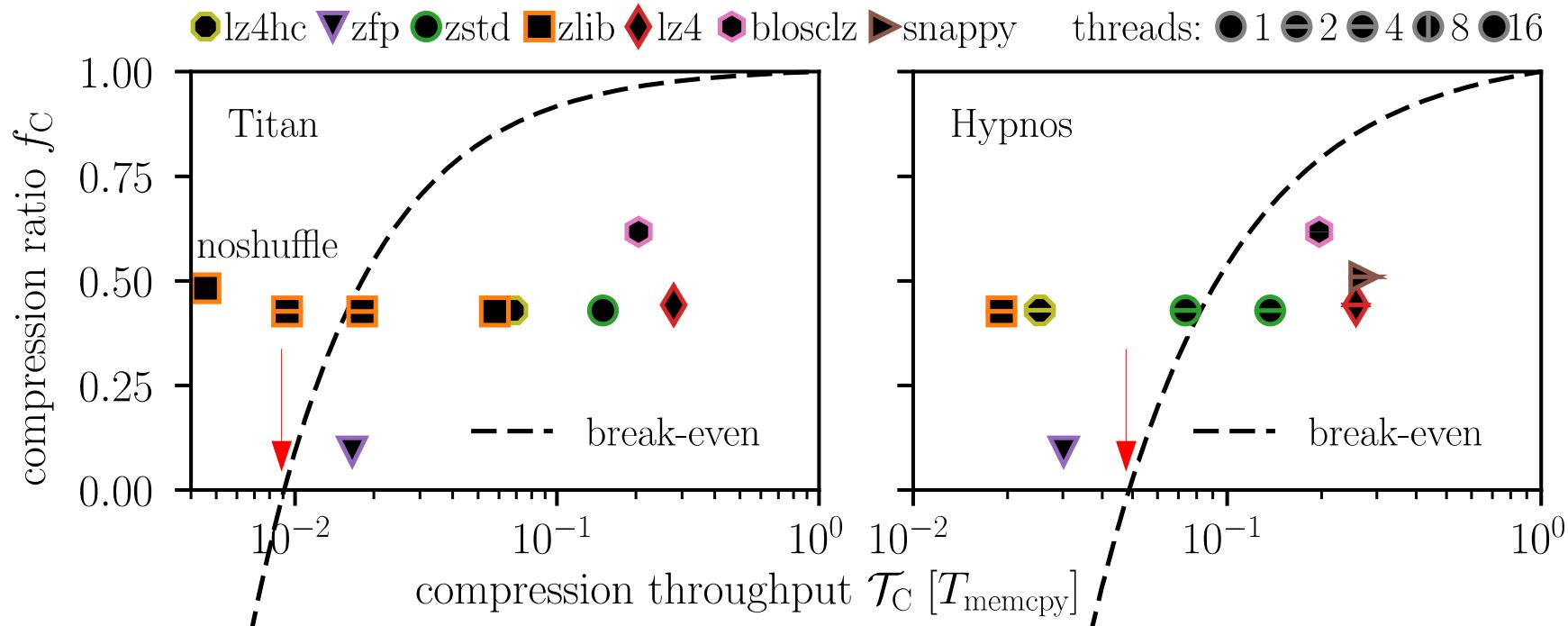
Which to choose??

Potential Solutions

and first steps taken

$$\mathcal{T}_C > \frac{\mathcal{T}_{FS}}{\mathcal{T}_{FS} + 1}$$

On-the-fly (Lossless) Compression



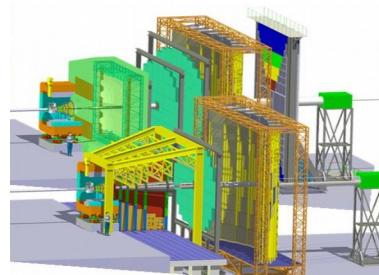
Zfp 0.5.1: three uncompressed bits / scalar; on *particle data*

Blosc 1.11.4-dev: add bitshuffle pre-conditioner

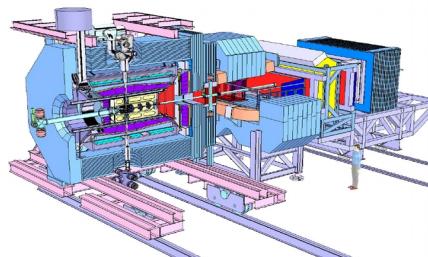
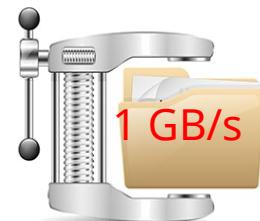
In Transit Data Reduction Blurring Online and Offline by ALICE and FAIR Experiments



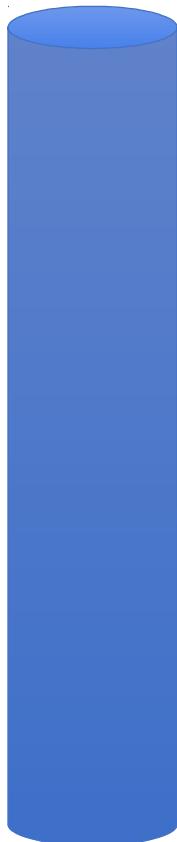
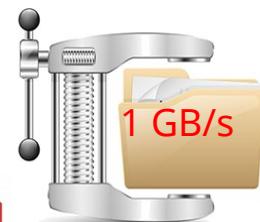
3.4 TB/s



1 TB/s



0.3 TB/s

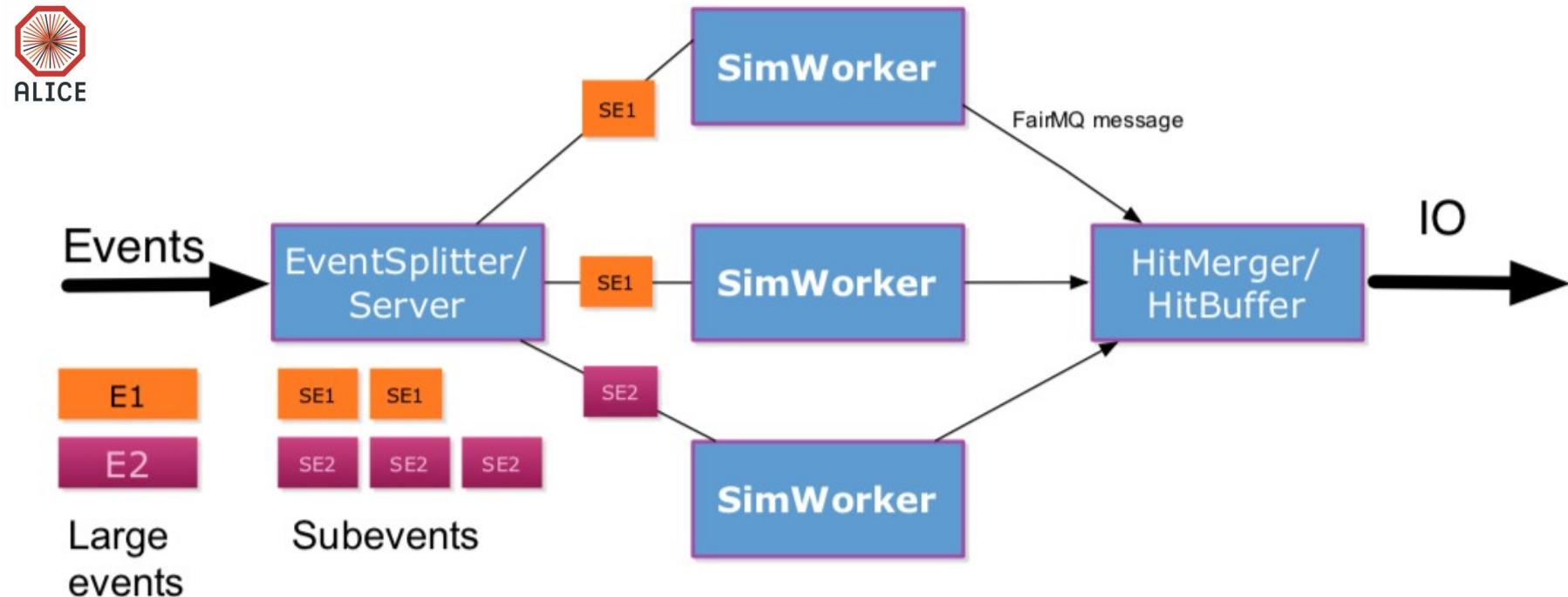


The unifying element: Detector Readout

- Continuous readout with self-triggered front-end electronics
- No or limited hardware triggers
- Event definition & selection requires full reconstruction in online compute farms

In Transit Data Processing

E.g. Using FairMQ



Human In-the-Loop In-Situ Data Reduction

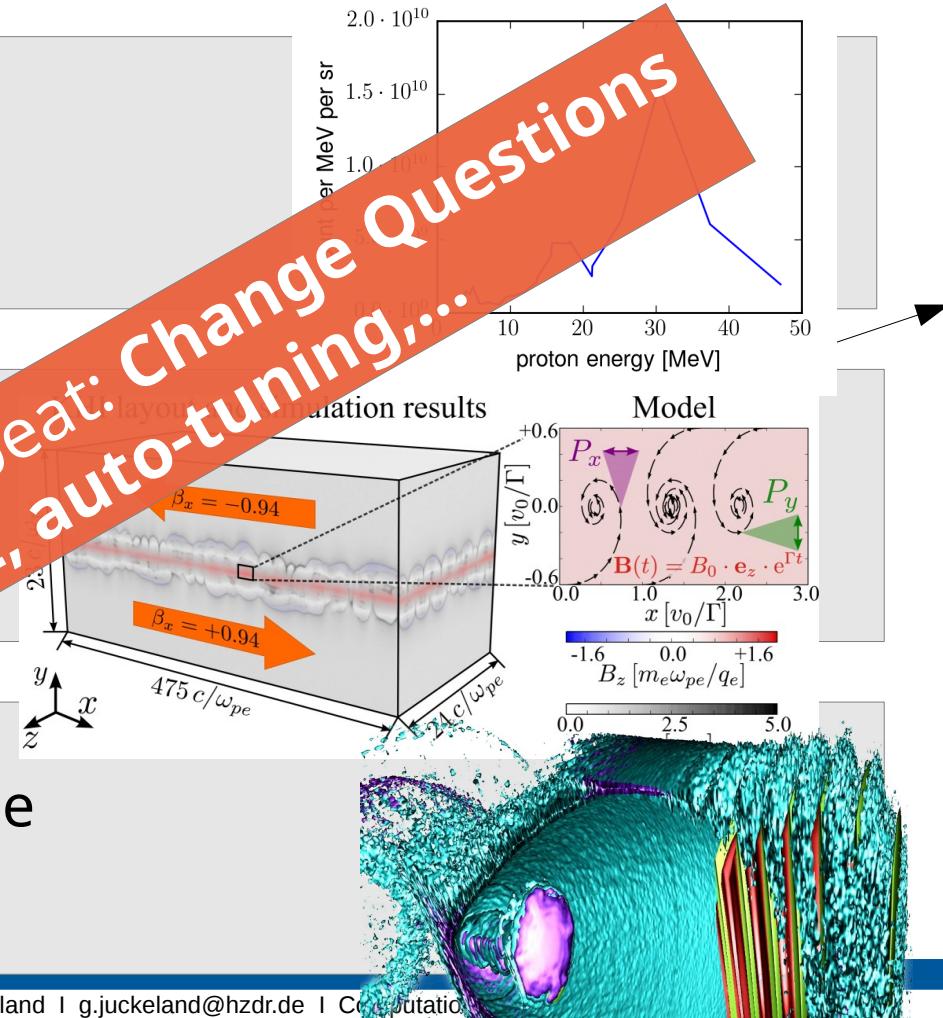
A. Huebl et al. (2014), DOI:10.1109/TPS.2014.2327392
R. Pausch et al. (2017), DOI:10.1103/PhysRevE.96.013316
A. Matthes, A. Huebl et al., ISC'16 (2016), DOI:10.14529/jsfi160403
A. Huebl et al., ISC'17 (2017), DOI:10.1007/978-3-319-67630-2_2

Binning of a **spectrogram**
Creation of a **phase space image**

In situ **radiation diagnostic**

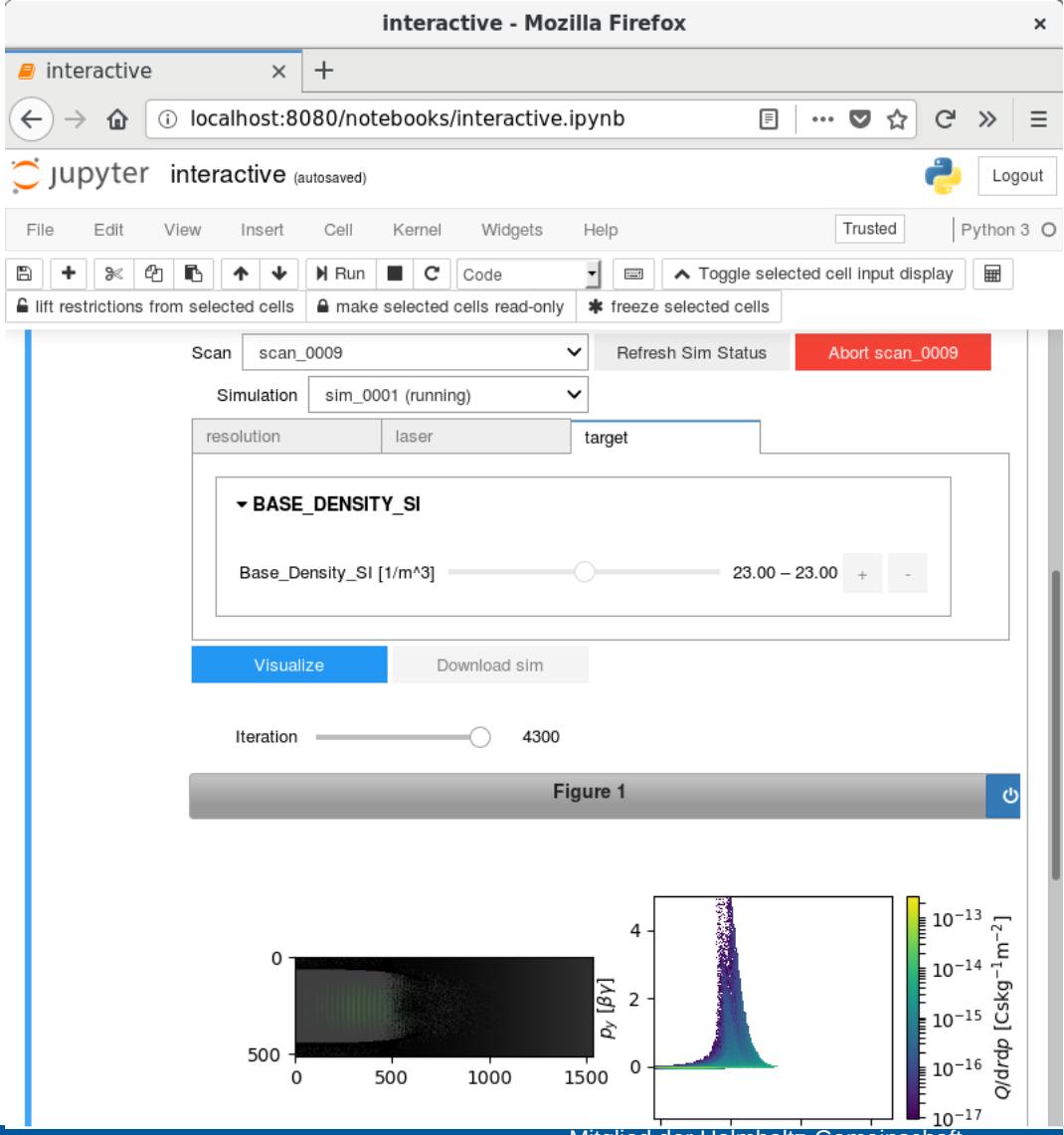
Ray-cast
Lossy data compression

Observe, Correlate & Repeat: Change Questions
Later: Add ML, auto-tuning....



Data-Driven, Interactive Exploration

- **data** (reduced or not)
open meta standard
- **scientific questions**
 - python & GUI scripts
 - transparent (hackable)
- **Jupyter integration**
- **installability**
 - containerization
 - package manager



Interactive JIT CUDA/C++/...

jupyter CUDA_copy (autosaved) Logout

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

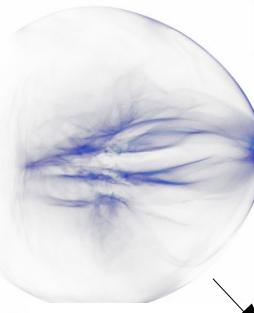
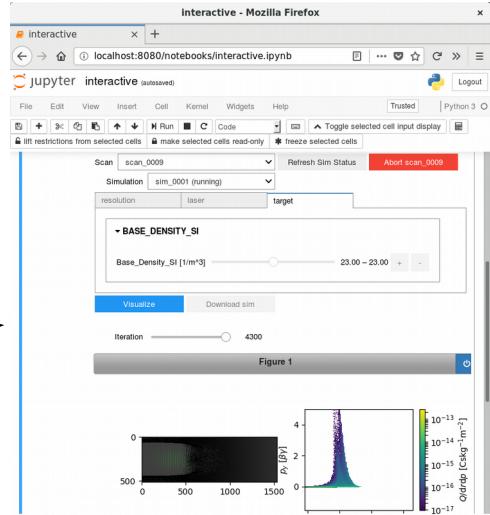
In []: `template <typename T>
__global__ void copy_kernel(T * in, T * out, unsigned int N){
 int id = blockIdx.x * blockDim.x + threadIdx.x;
 if(id < N)
 out[id] = in[id];
}`



our cling contribution :)

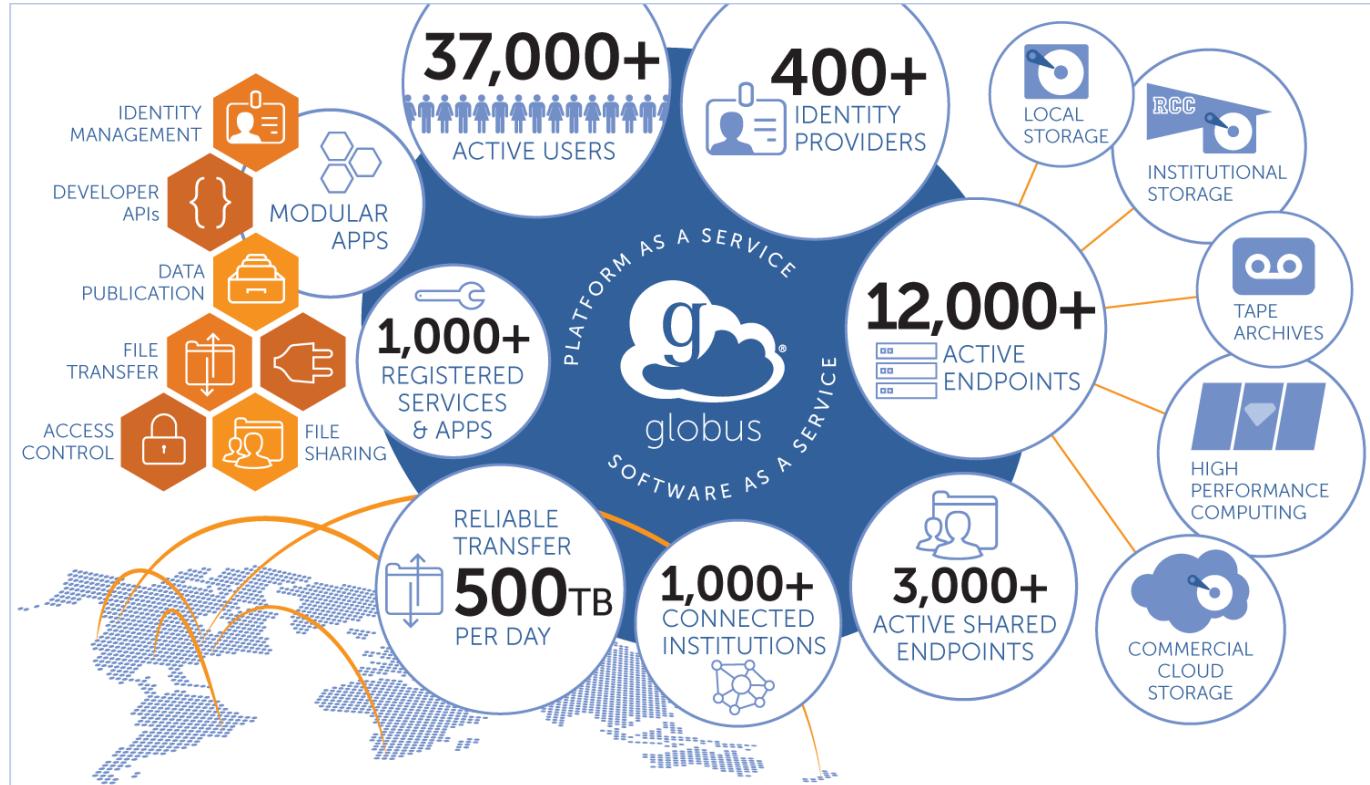
Cling CUDA: S. Ehrig (HZDR, TU Dresden), Diploma Thesis
(2018)

Bringing it all together



Surprise: Others call it
Cloud Computing

Data Transfers between Data Centers



Summary + Outlook

- You want to reduce data (not blindly compress)
- It's hard to know the right <0.01% beforehand in explorative science (bleeding-edge experiments and simulations)
- The data will come (sooner than later)
- Keep data „in-flight“
- In-situ techniques (with or without a human) help producing „good“ data sets
- Use same software stack for simulation and experiment evaluation
- **Moving data is expensive** – do it at least asynchronously