

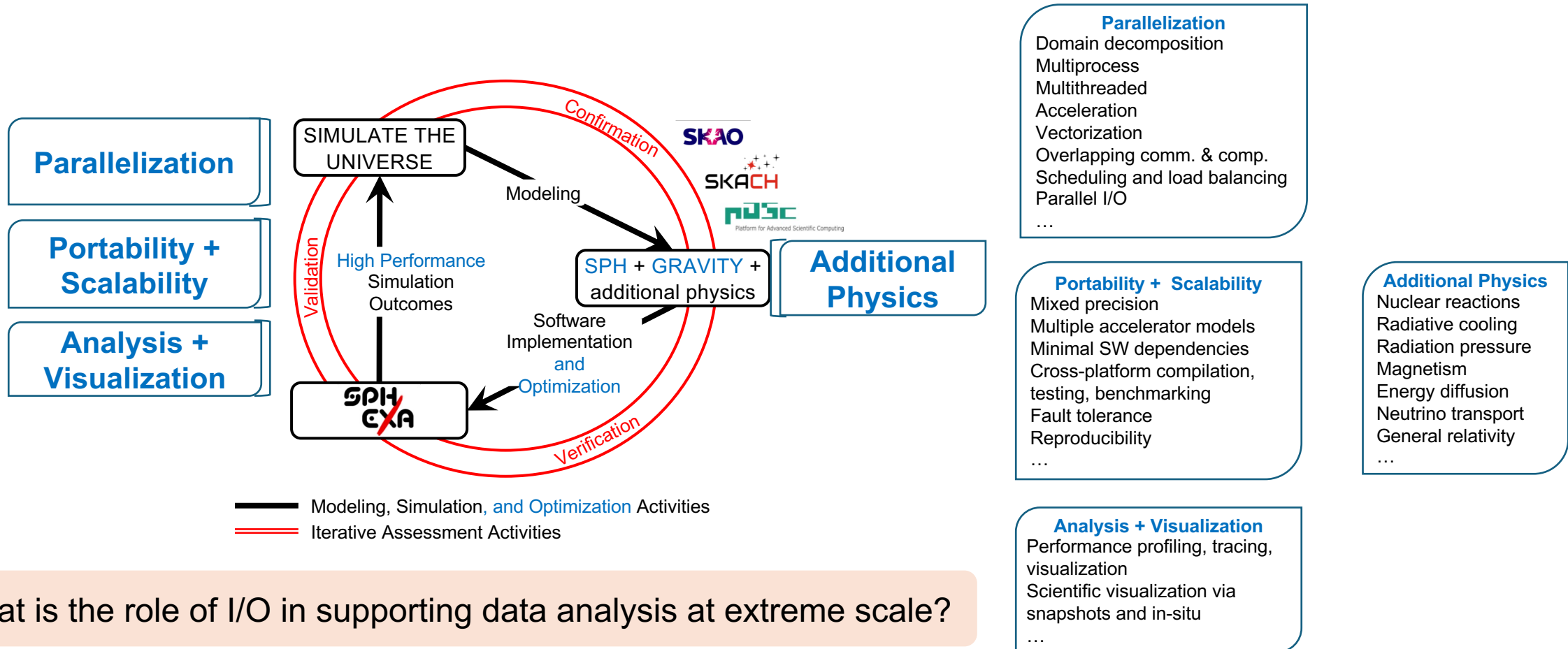
Compression for Extreme-Scale Computing: A Case Study of SPH-EXA's Data Analysis Architecture

Yiqing Zhu, Osman Seckin Simsek, Rubén Cabezón, Florina Ciorba
University of Basel

SKACH Winter Meeting, January 22, 2024



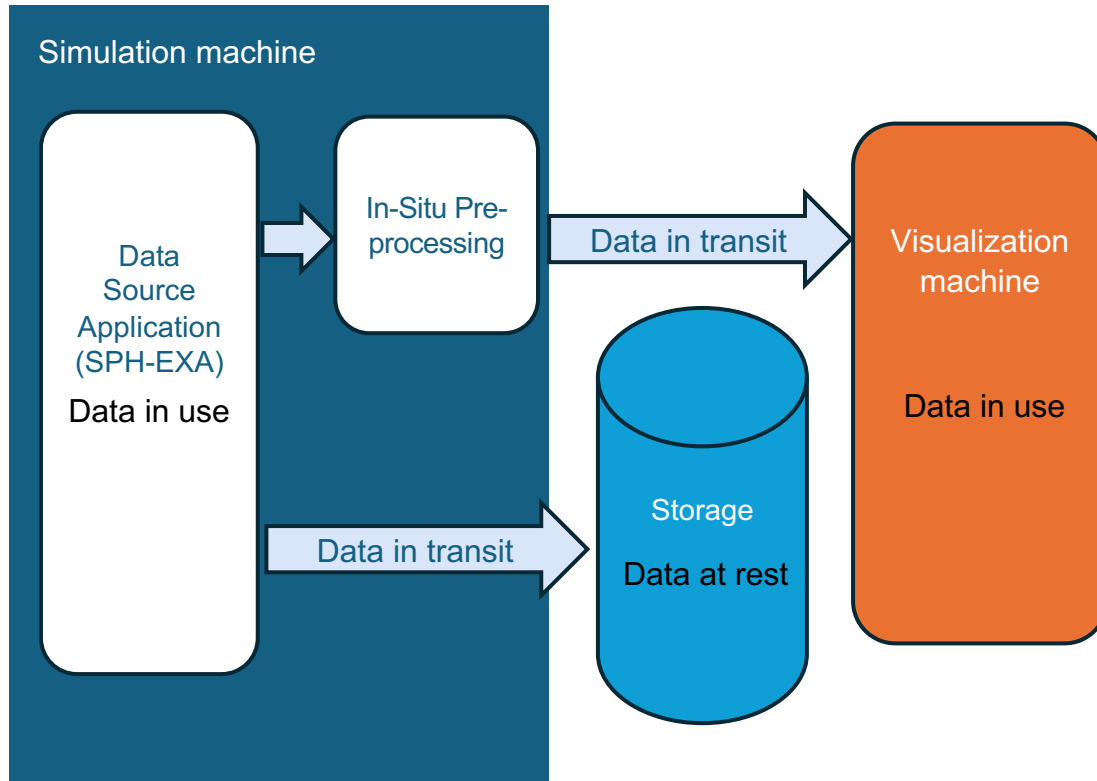
Modeling, Simulation and Optimization through Interdisciplinary Co-Design



What is the role of I/O in supporting data analysis at extreme scale?

[Adapted from: Schlesinger, S., "Terminology for Model Credibility," Simulation, Vol. 32, No. 3, 1979.]

Extreme-Scale Data Analysis: Post-Hoc and In-Situ



SPH-EXA data analysis architecture: Post-hoc and in-situ

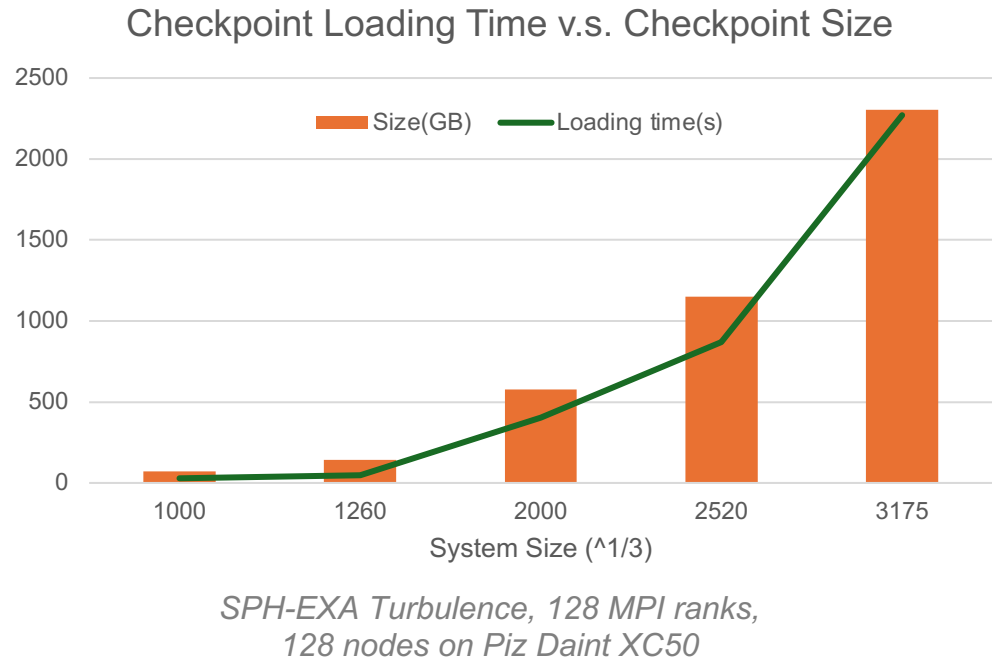
Post-hoc Analysis

- From simulation machine directly output to storage or visualization pipeline
- High I/O
- Needs a powerful visualization machine

In-situ Analysis

- Runtime co-process the simulation data, then outputs to visualization pipeline
- Can extract features in runtime
- Low I/O
- Low HW requirement (visualization can even be in-place)

Extreme-Scale Data Analysis: Post-Hoc and In-Situ



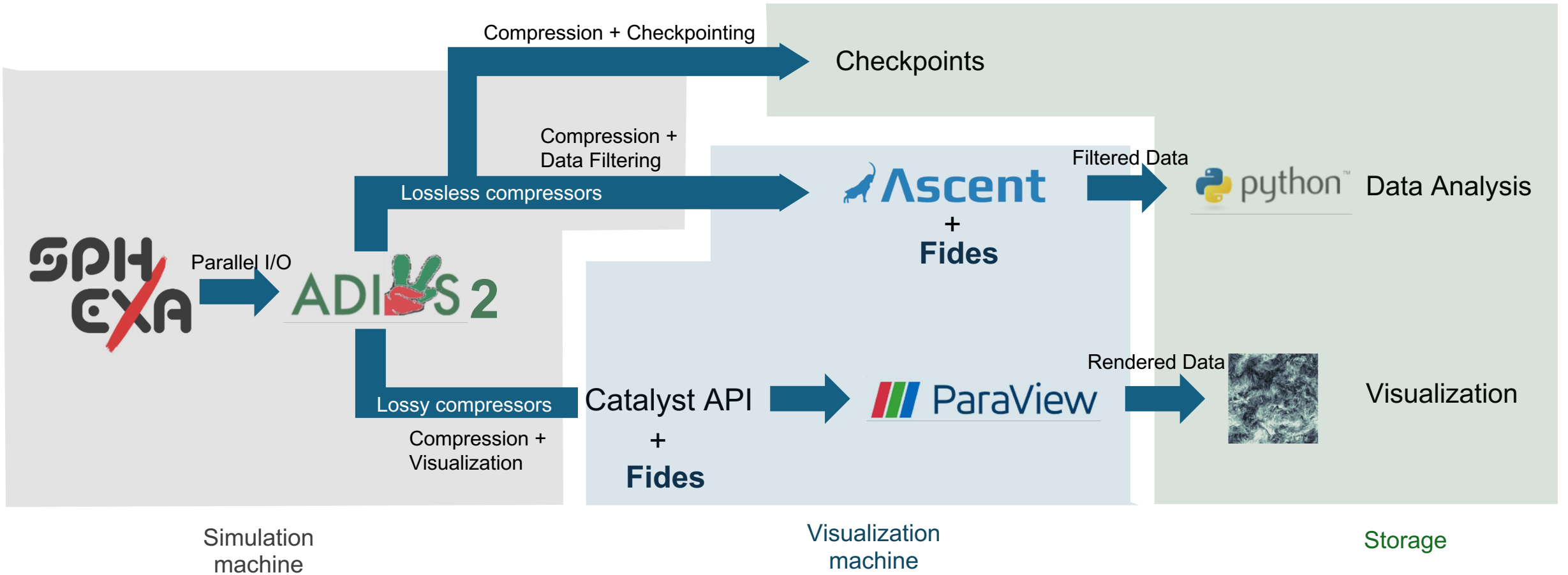
Challenges

- Reduce checkpointing time during simulation
- In-situ process of large datasets
- Parallel non-blocking I/O

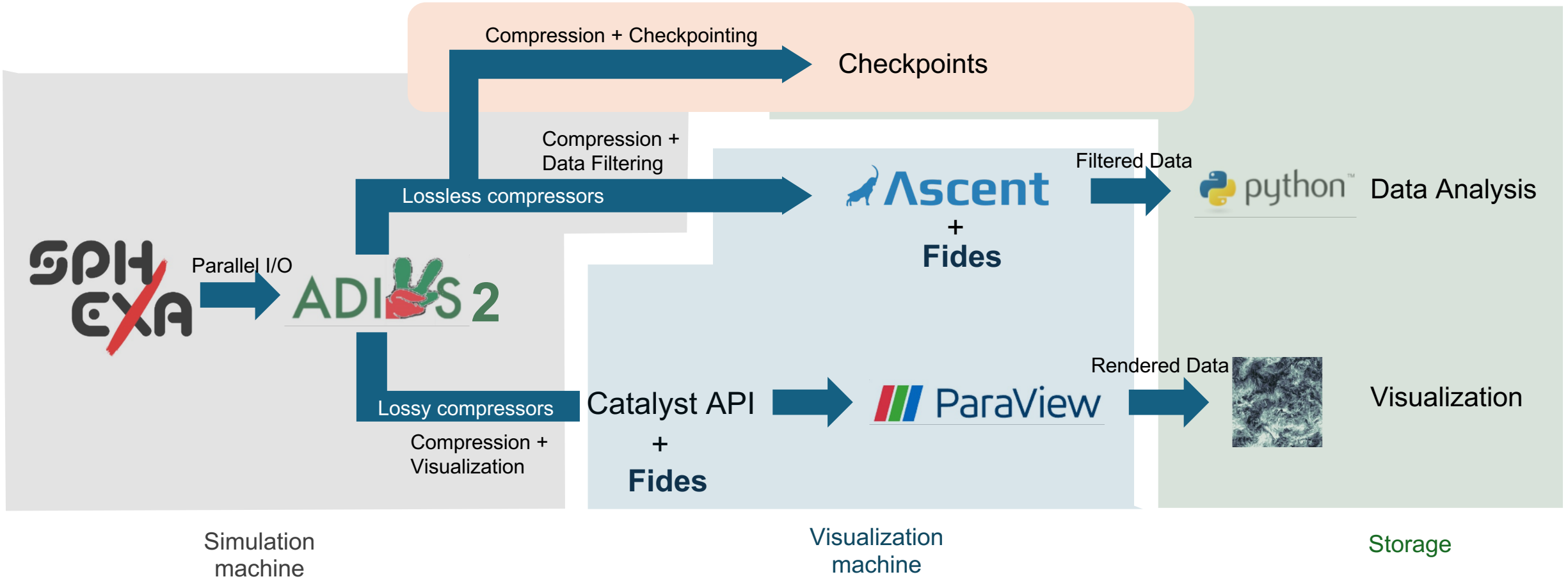
Possible Solution: Compression

Data storage, loading, transfer...

SPH-EXA Checkpointing and Data Analysis



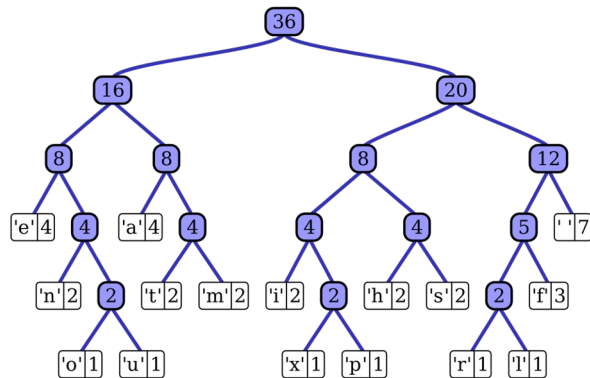
Checkpointing: Lossless Compression



Checkpointing: Lossless Compression

- Lossless compressors

- Zlib
- LZ
- LZ4
- LZ4HC



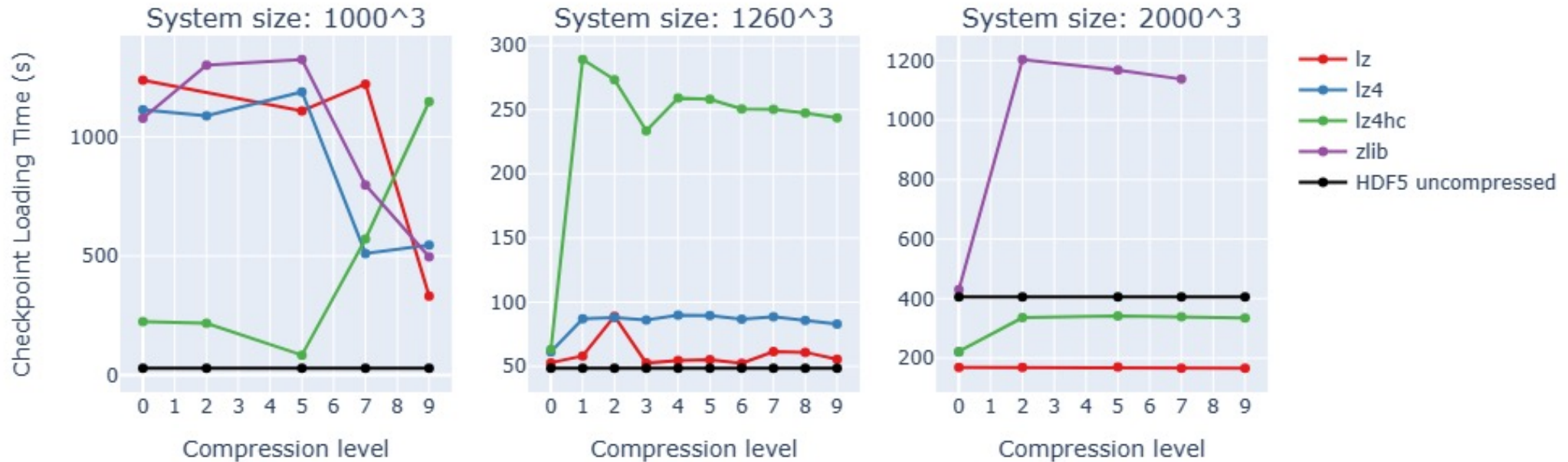
[Meteficha. (2023, December 2). Huffman coding. Wikipedia. https://en.wikipedia.org/wiki/Huffman_coding.]



SPH-EXA Turbulence, 128 MPI ranks, 128 nodes on Piz Daint XC50

Checkpointing: Lossless Compression

Checkpoint Loading Time w.r.t. Compression Level



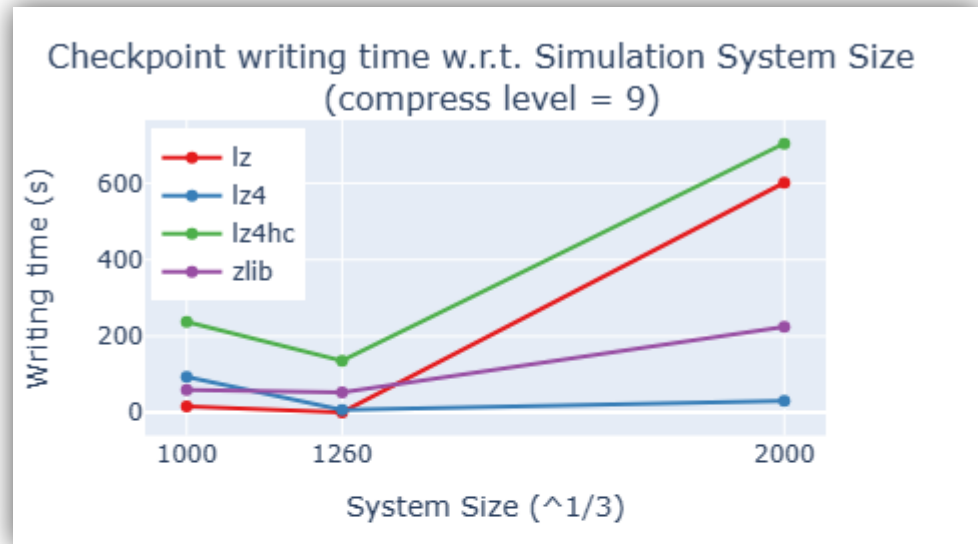
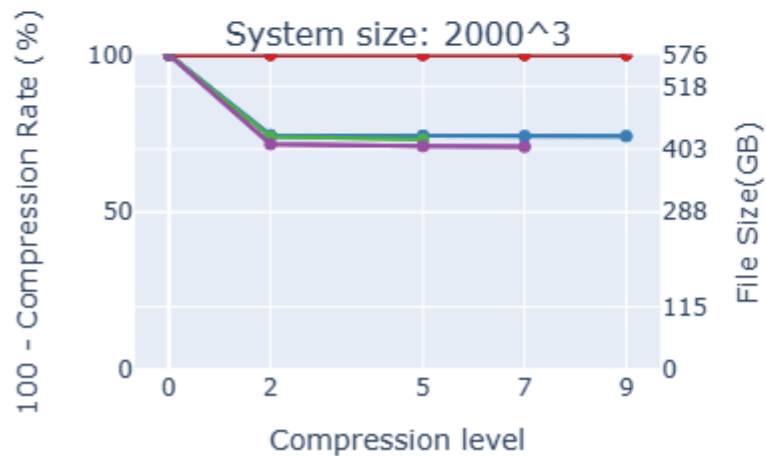
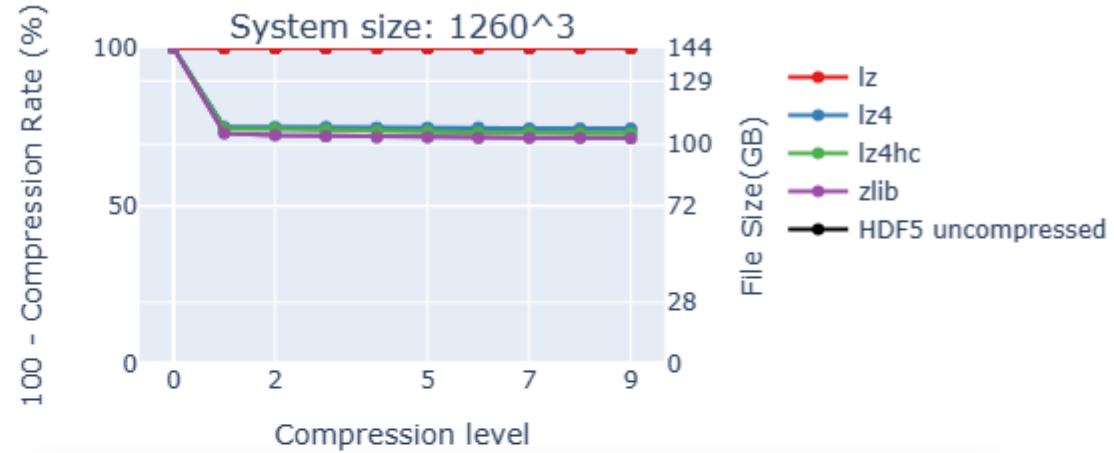
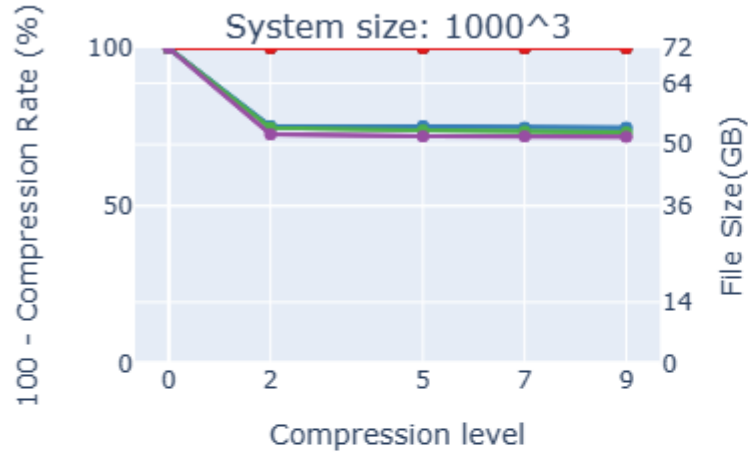
SPH-EXA Turbulence, 128 MPI ranks, 128 nodes on Piz Daint XC50

For most lossless compressors, decompression cost decreases with compression level

Checkpointing: Lossless Compression

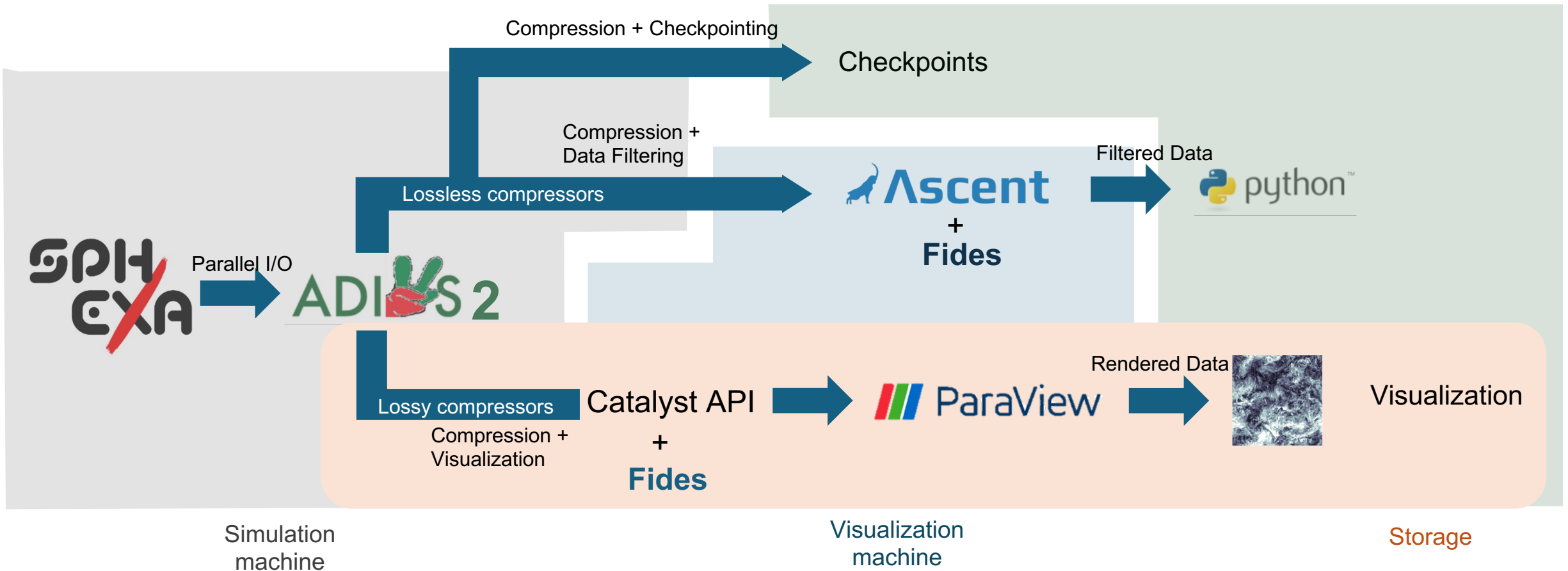
$$\text{Compression rate: } 1 - \frac{\text{File Size}_{\text{compressed}}}{\text{File Size}_{\text{uncompressed}}}$$

Lossless Compression Rate w.r.t. Compression Level



SPH-EXA
Turbulence, 128
MPI ranks, 128
nodes on Piz Daint
XC50

Visualization: Lossy Compression

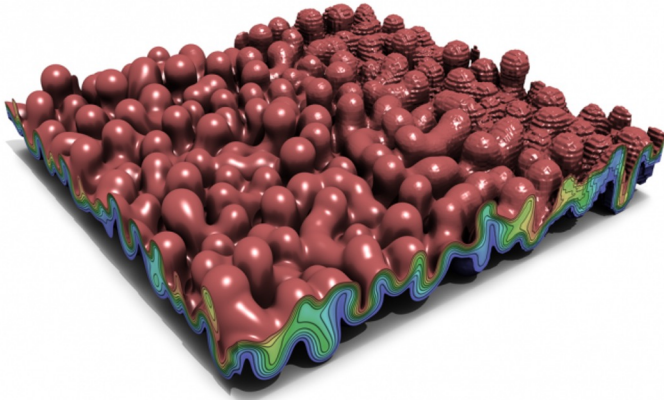


Visualization: Lossy Compression

- Lossy Compressors

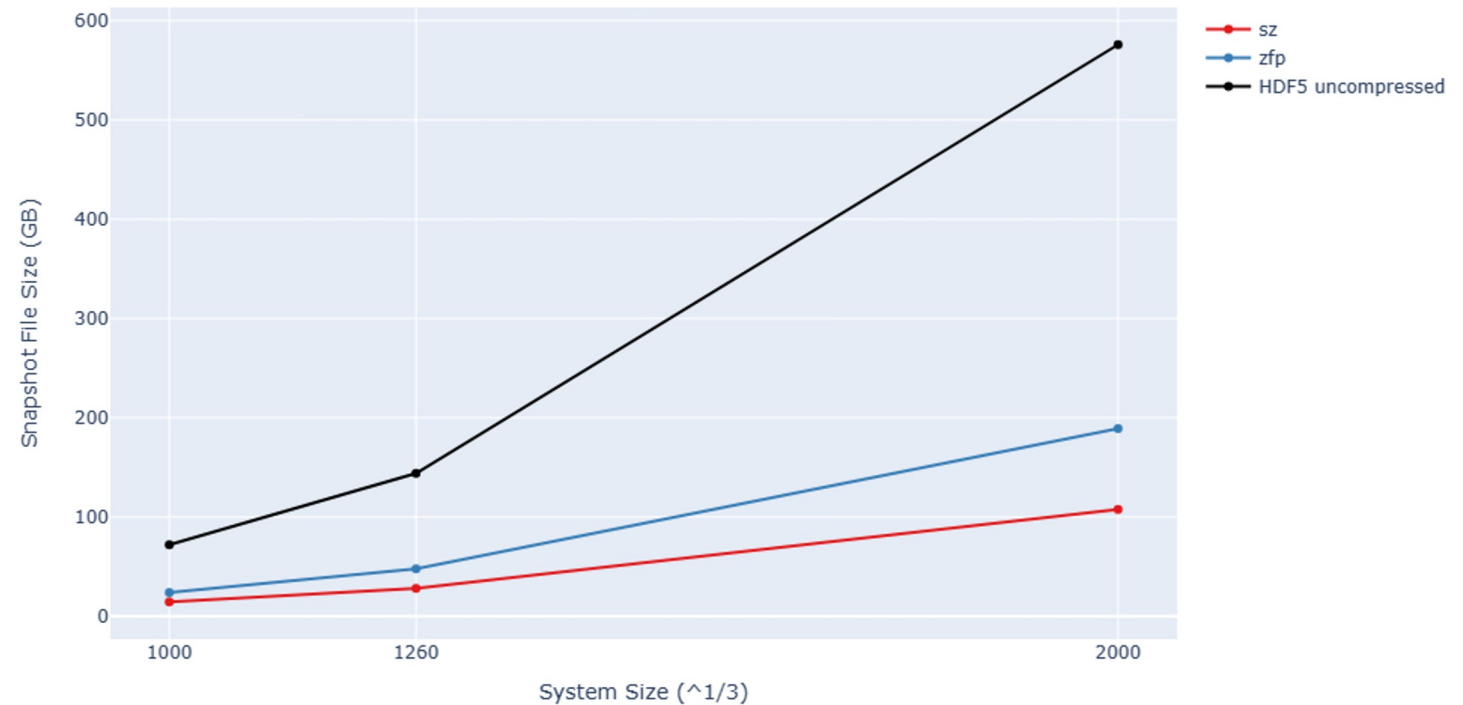
- SZ
- ZFP

Given uncompressed value f and reconstructed value g , it is guaranteed that $|f - g| \leq \text{accuracy}$



[Lindstrom, P. (2014). Fixed-rate compressed floating-point arrays. *IEEE transactions on visualization and computer graphics*, 20(12), 2674-2683.]

Lossy Snapshot File Size w.r.t. Simulation scale (accuracy= 10^{-5})

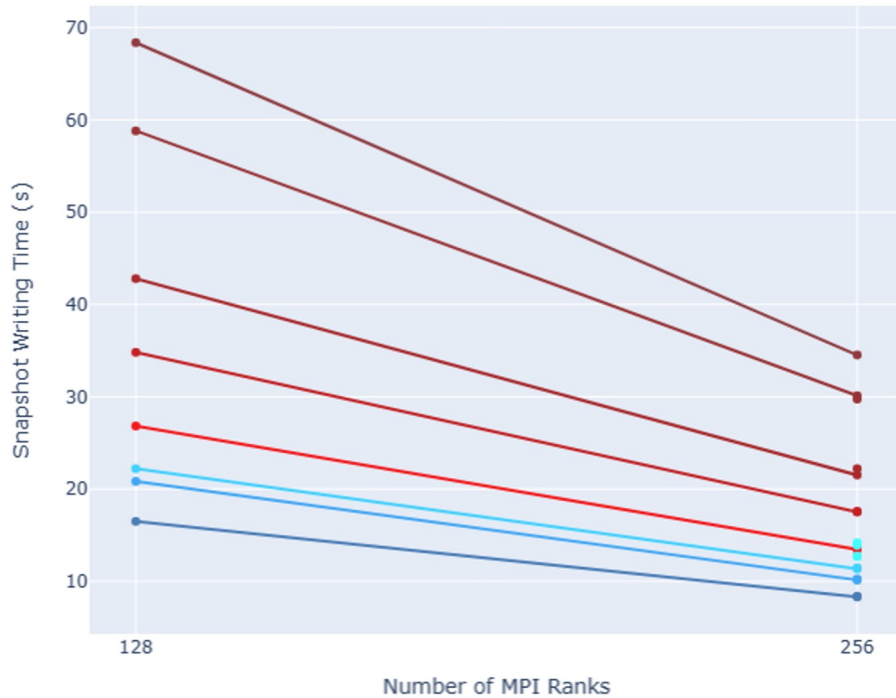


SPH-EXA Turbulence, 128 MPI ranks, 128 nodes on Piz Daint XC50

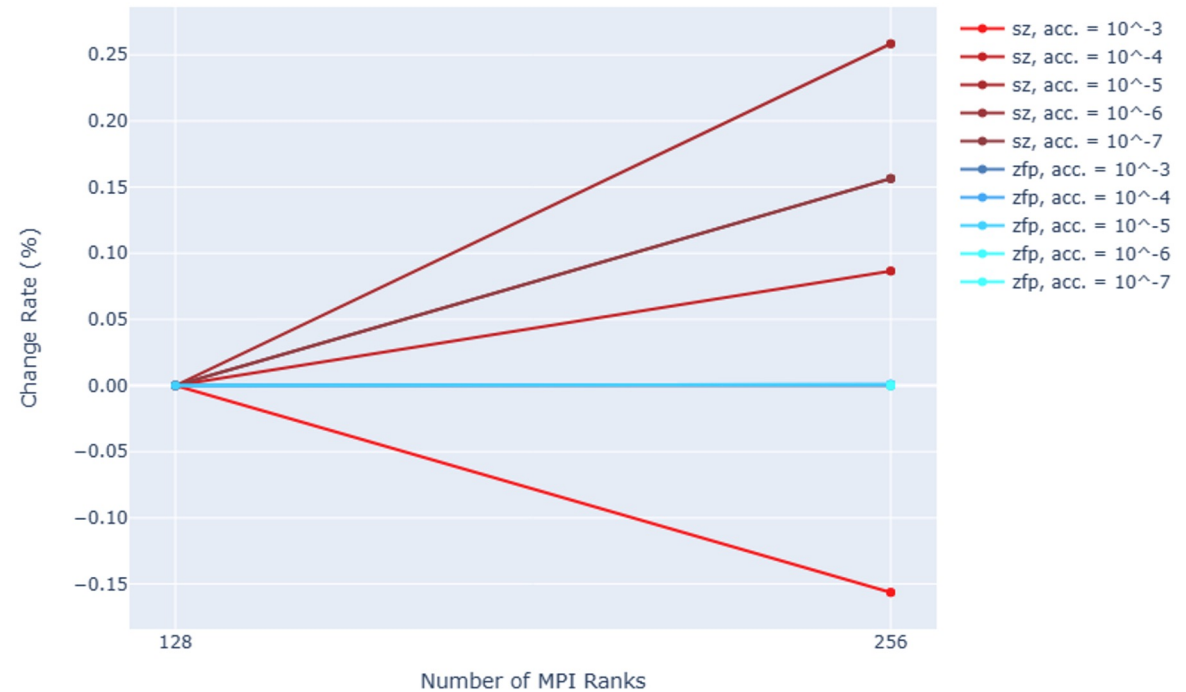
Visualization: Lossy Compression

$$\text{Snapshot Size Change Rate} = \frac{\text{File Size}_{256 \text{ Ranks}}}{\text{File Size}_{128 \text{ Ranks}}} - 1$$

Lossy Compressors: Snapshot Writing Time vs. Number of MPI Ranks (system size = 1260³)



Lossy Compressors: Snapshot Size Change Rate w.r.t. Number of MPI Ranks (System size = 1260³)



SPH-EXA Turbulence, 128 MPI ranks, 128 nodes on Piz Daint XC50

Visualization: Lossy Compression

Compressor: SZ

Compressor: ZFP

*SPH-EXA
Turbulence, system
size 1000^3 at time
 $t=18.0s$. Colored by
velocity, with different
lossy compression
accuracy*

Zhu, Simsek,
Cabezon, Ciorba

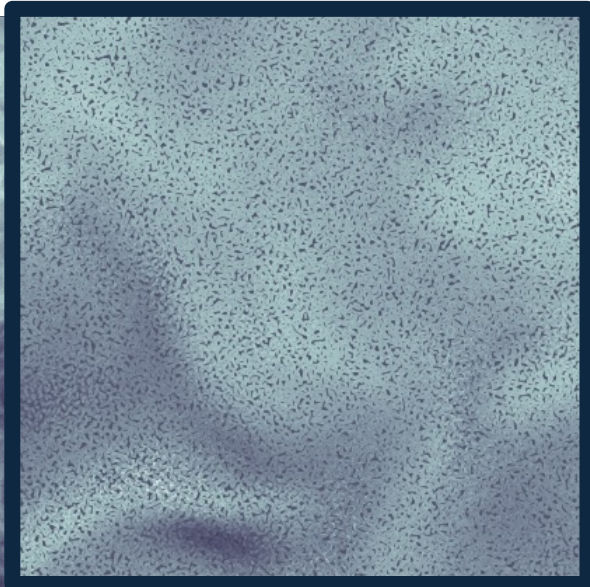
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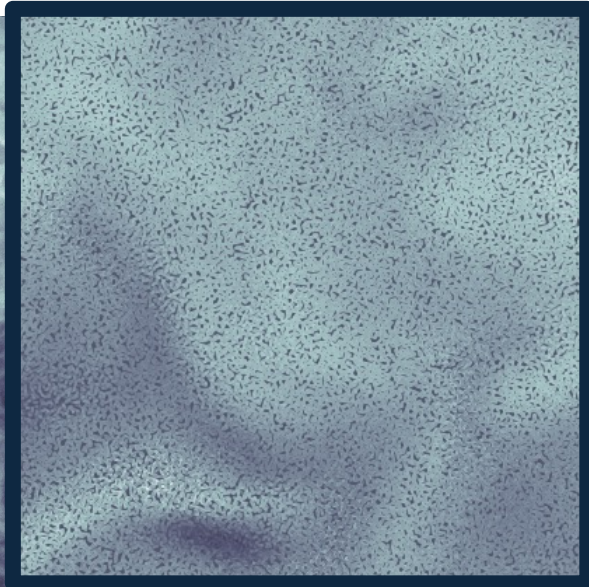
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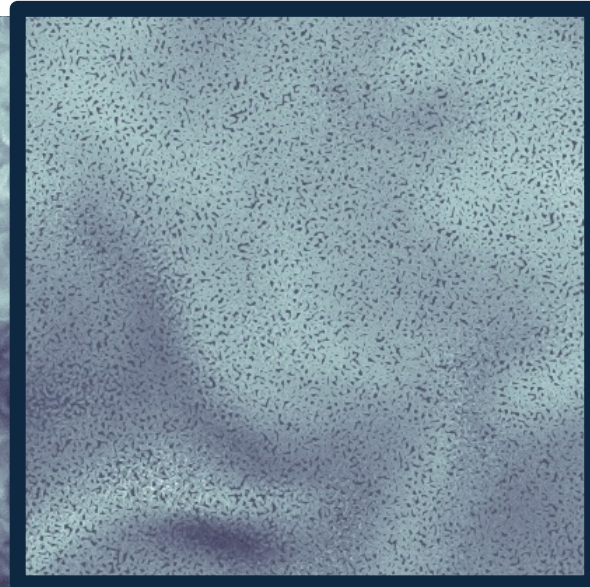
Visualization: Lossy Compression



SZ Accuracy= 10^{-4}

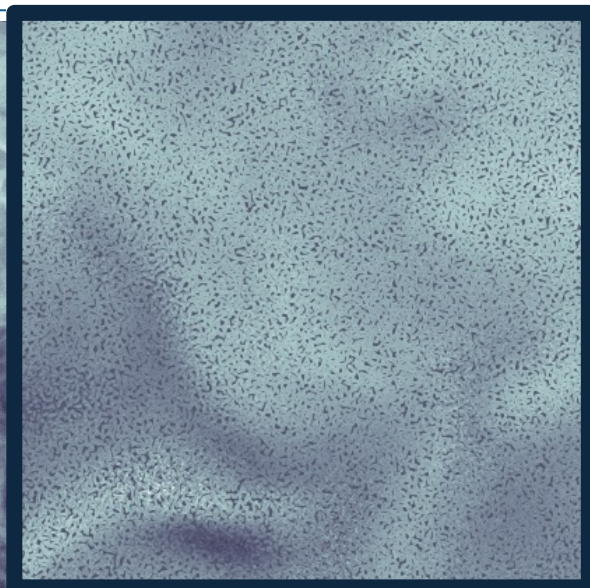


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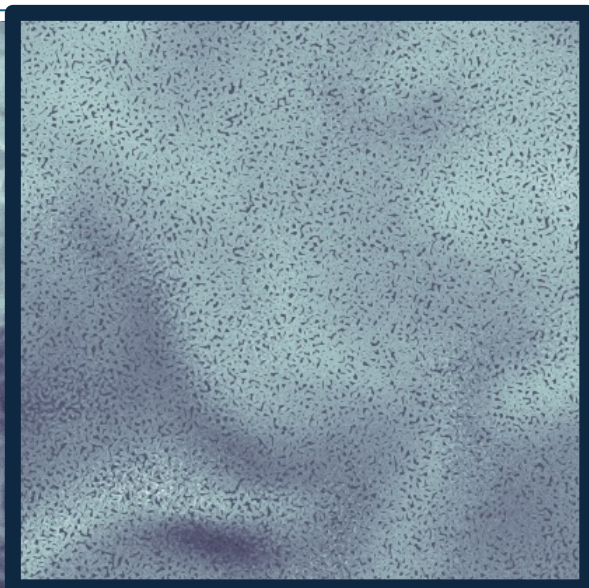


SZ Accuracy= 10^{-6}

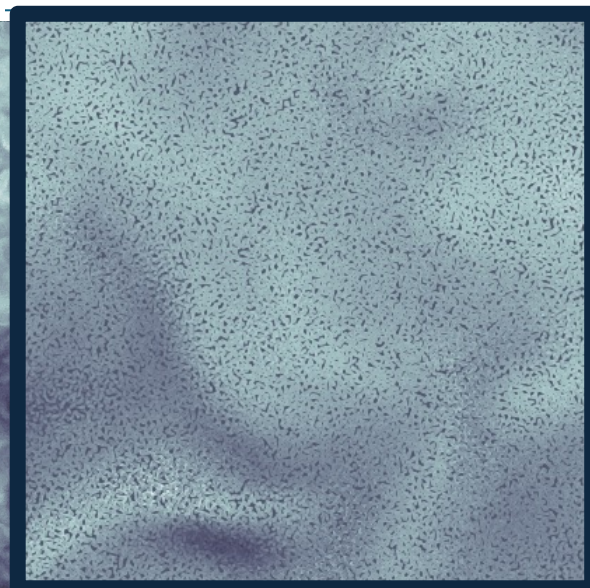
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ZFP Accuracy= 10^{-4}



ZFP Accuracy= 10^{-5}

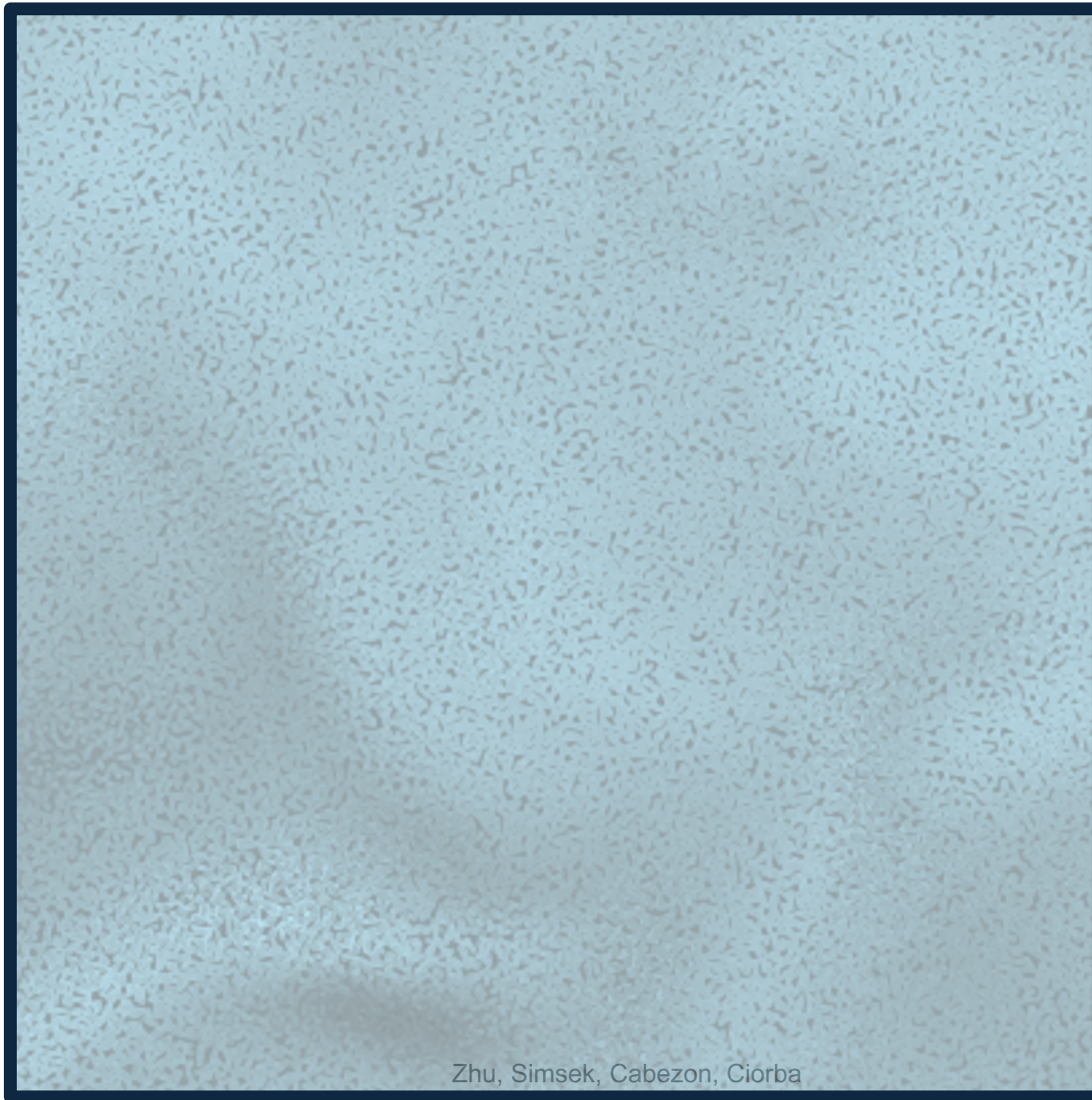


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Compressor: ZFP

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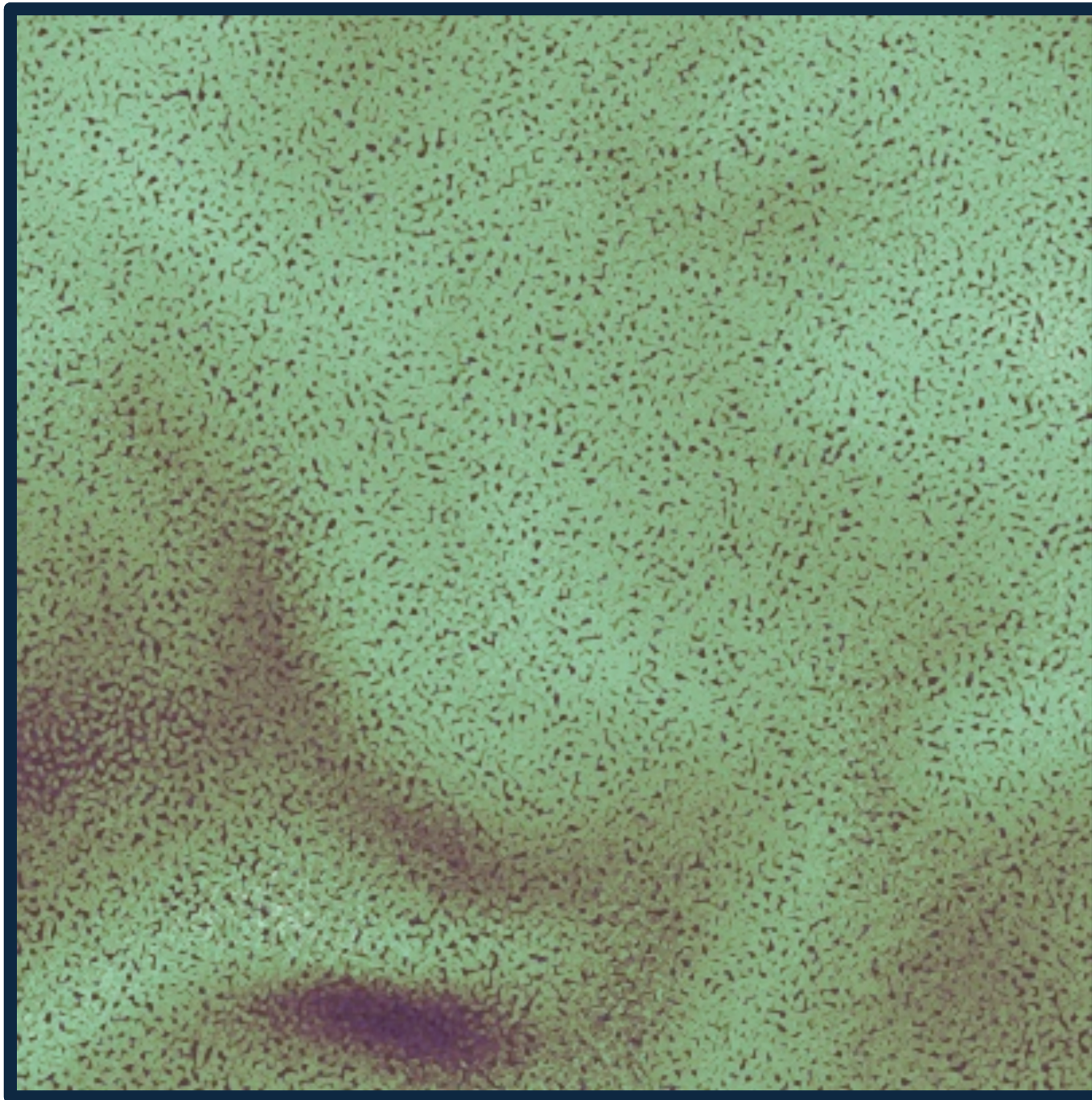
Visualization: Lossy Compression



SZ accu. = 10^{-4} (yellow)
SZ accu. = 10^{-6} (blue)

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Summary

- Exploration of compression for checkpointing
- Lossy compression for in-situ visualization
- Provide reference parameters for tuning compressors for extreme-scale simulations

What if we have higher I/O bandwidth...?

In Progress

- Mixed-precision
- Larger scale visualization
- Verification of compression
- ...

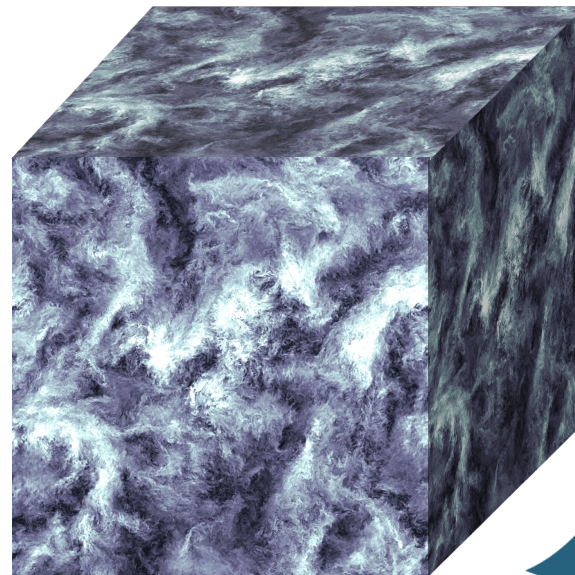
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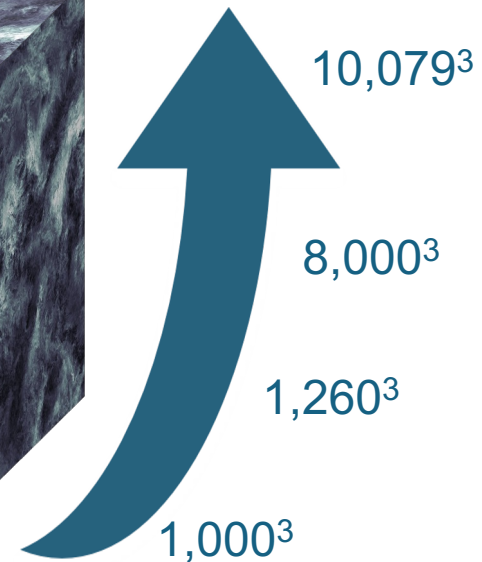
Projects We Support

TGSF: The Role of Turbulence and Gravity in Star Formation, Unveiling the Sonic Scale with Smoothed Particle Hydrodynamics



In Progress

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- ...



For 10,079³ particles, relaxing the ICs already consumes 500'000 node hours on LUMI-G

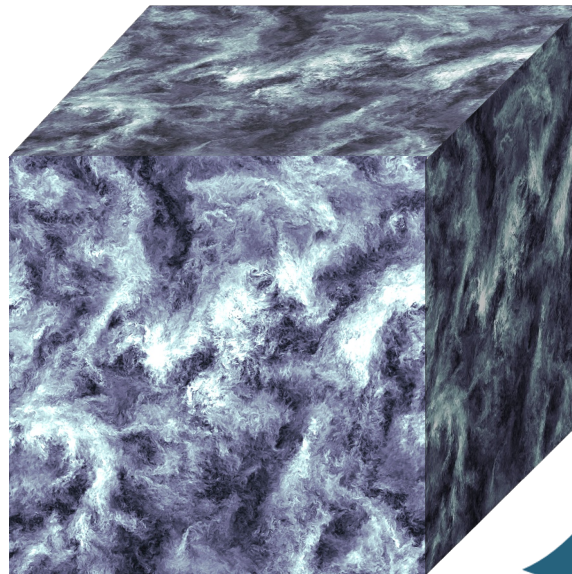
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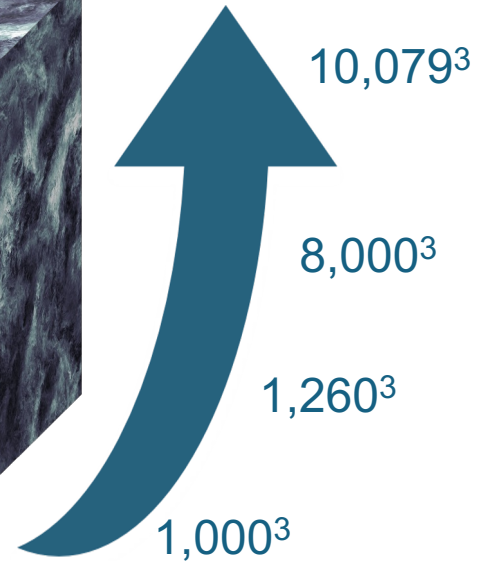
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In Progress

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Take away

We explore the potential of an efficient data analysis architecture to support extreme-scale simulations, such that storage, data loading and data transfer are not a bottleneck.

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