

Chair for High Performance Computing
Philipp Neumann

TaLPas: Task-Based Load Balancing and Auto-Tuning in Particle Simulations

TU München	Helmut- Schmidt Uni Uni Hamburg	HLRS	VISUS/ Uni Stuttgart	TU Darmstadt	TU Berlin	TU Kaiserslautern
H.-J. Bungartz	P. Neumann	C. Glass	G. Reina	F. Wolf	J. Vrabec	H. Hasse
N. Tchipev	T. Ludwig	J. Gracia	O. Fernandes	S. Shudler	M. Heinen	M. Horsch
S. Seckler		N. Urmersbach	P. Gralka	S. Rinke		K. Langenbach
F. Gratl		C. Niethammer				T. Vo
						M. Heier

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Disclaimer!

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X,Y

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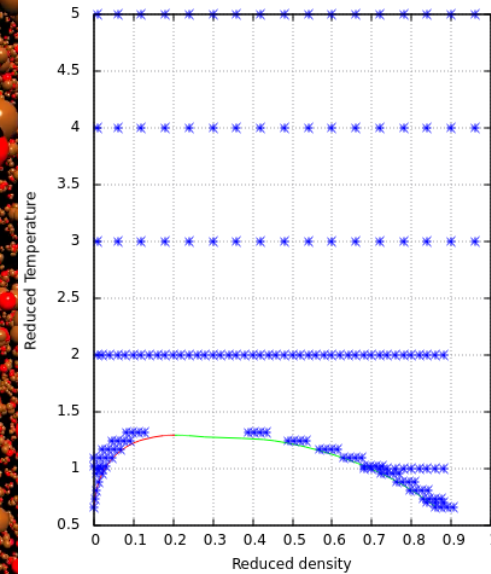
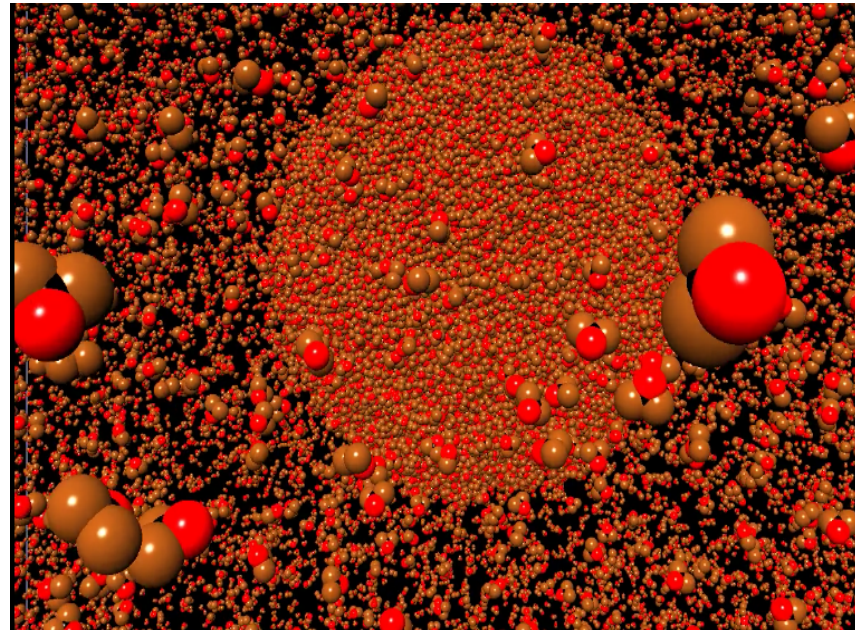
T. Vo

M. Heier



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Recap: What is TaLPas?



- Investigation of thermodynamic states and properties of fluids
→ vapor-liquid systems, interfacial flows, complex fluids, ...
- Particular computational challenge: problems such as equation of state sampling, rare events, ...
→ **many inter-dependent MD runs, each with different compute requirements**
→ Similar problem settings: UQ, parameter identification, ...

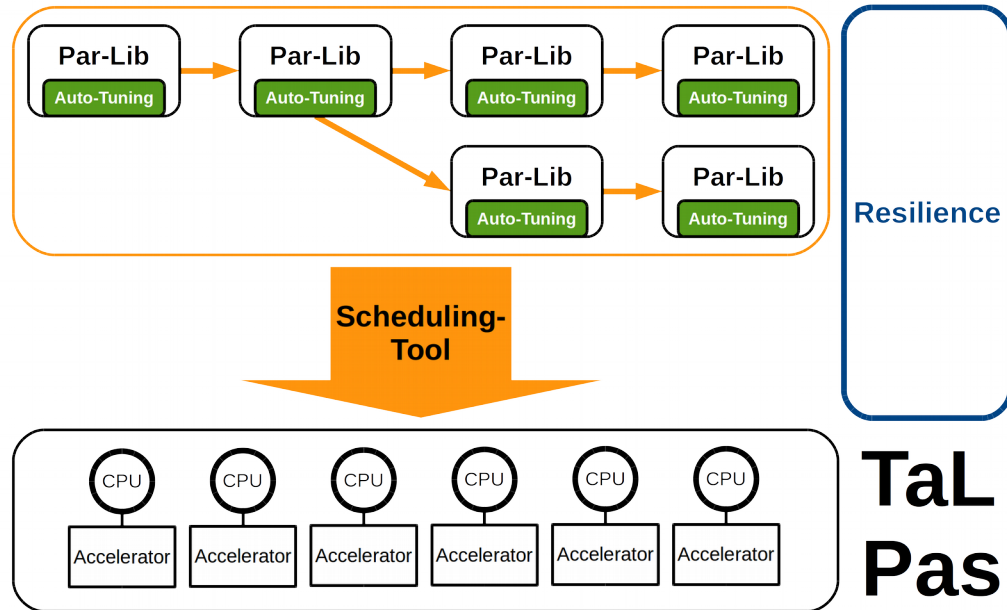
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Recap: Goals of TaLPas



- Hardware-independent acceleration of particle simulations
 - Node level auto-tuning library AutoPas
- Self-adapting performance-optimal distribution of work load
 - workflow manager, incorporating scheduler, performance predictor, particle sampling algorithm
- Resilience for particle systems
- Integrated (in-situ) visualization

Outline

- AutoPas
- Workflow Manager
- Performance Prediction
- In-Situ Visualization

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Outline

- 1 Auto-Tuning: The AutoPas Library
- 2 Workflow Manager and Performance Prediction
- 3 In-Situ Visualization
- 4 Summary

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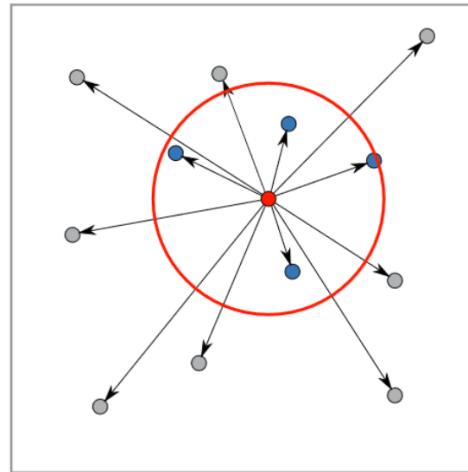


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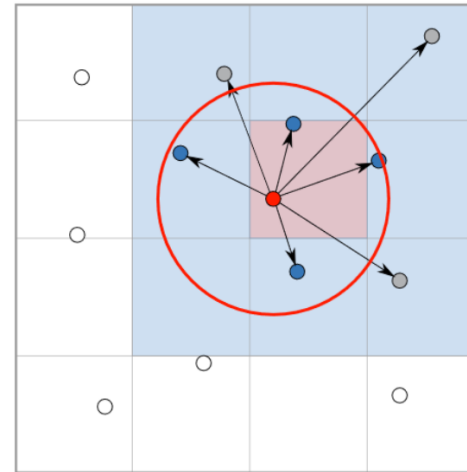
AutoPas¹

Outline

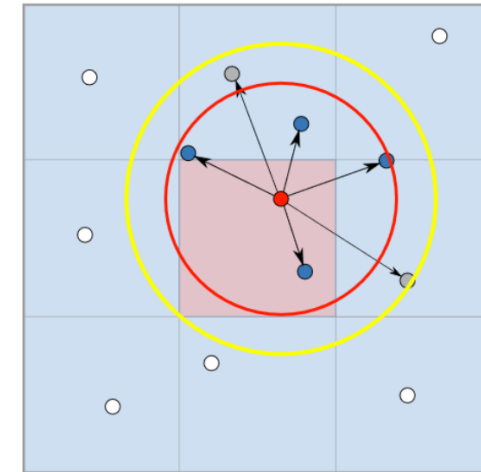
- AutoPas
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(a) DirectSum



(b) LinkedCells



(c) VerletLists

- Data structures: SoA vs. AoS
- 7 containers: Direct sum, linked cells, Verlet lists, cluster lists
- 17 (OpenMP) traversals: coloring, slicing, etc.
- Auto-Tuning on data structures+containers+traversals
- Integration in Is1 mardyn²

¹ Gratl et al. AutoPas: Auto-Tuning for Particle Simulations. IPDPS proc. (iWAPT workshop), 2019

² Tchipev et al. TweTriS: Twenty Trillion-atom Simulation. IJHPCA 33(5):838-854, 2019

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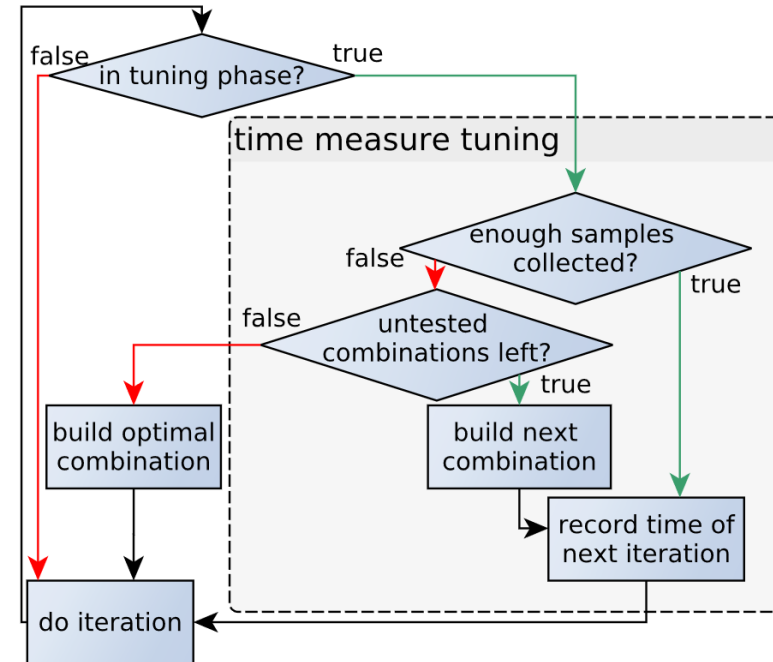
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AutoPas¹

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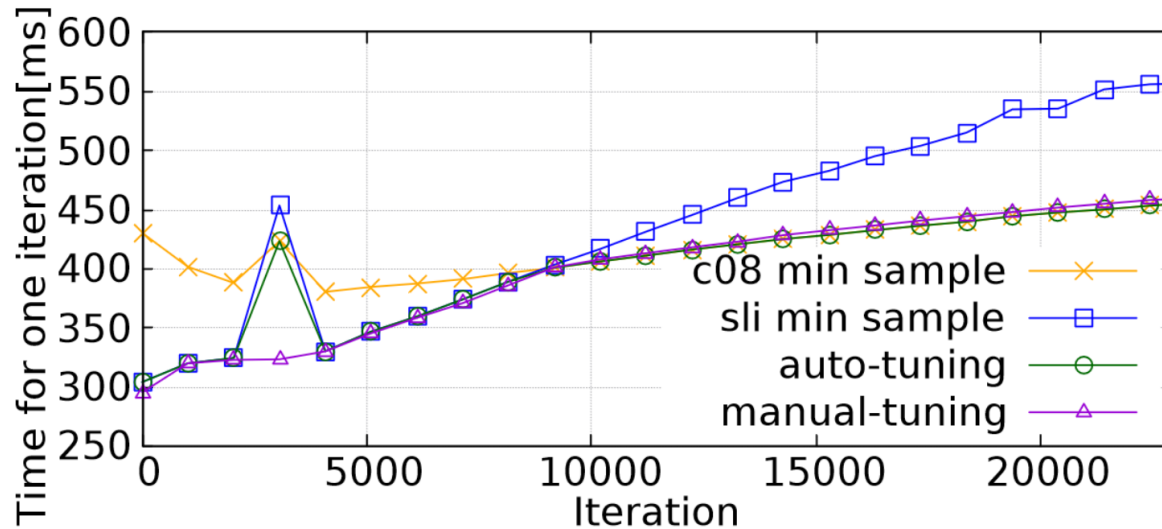


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AutoPas¹

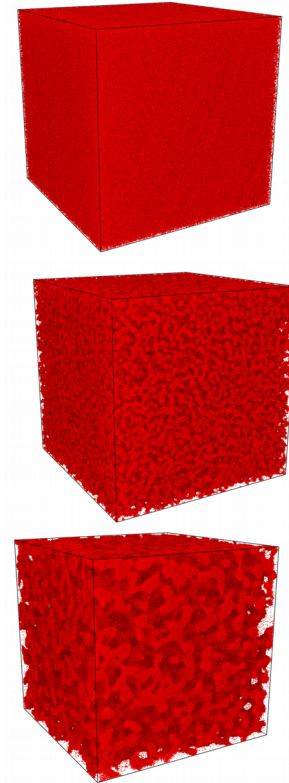


Number of Samples: 10

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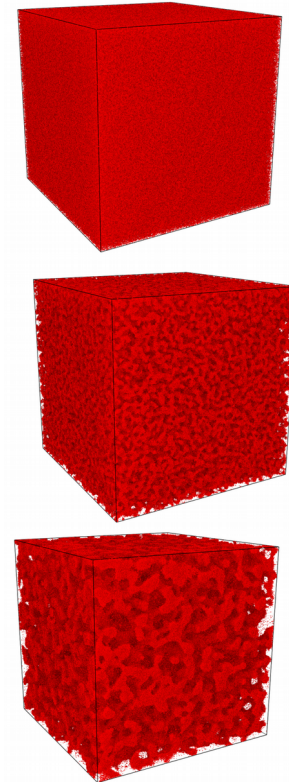
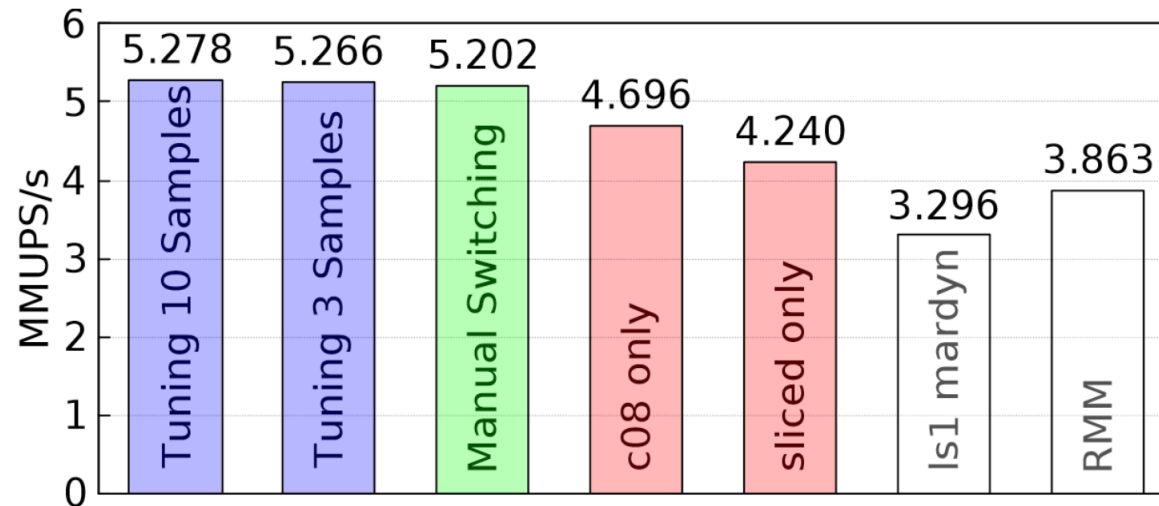


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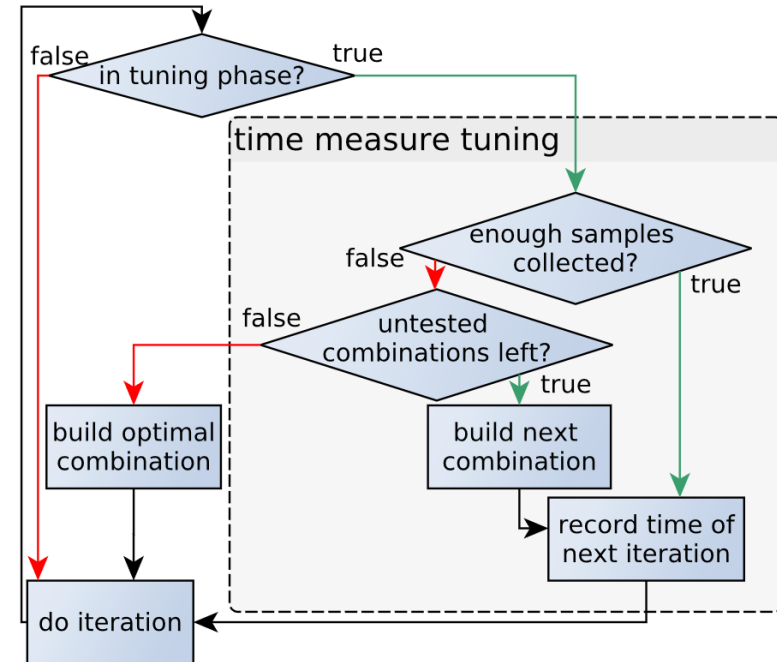


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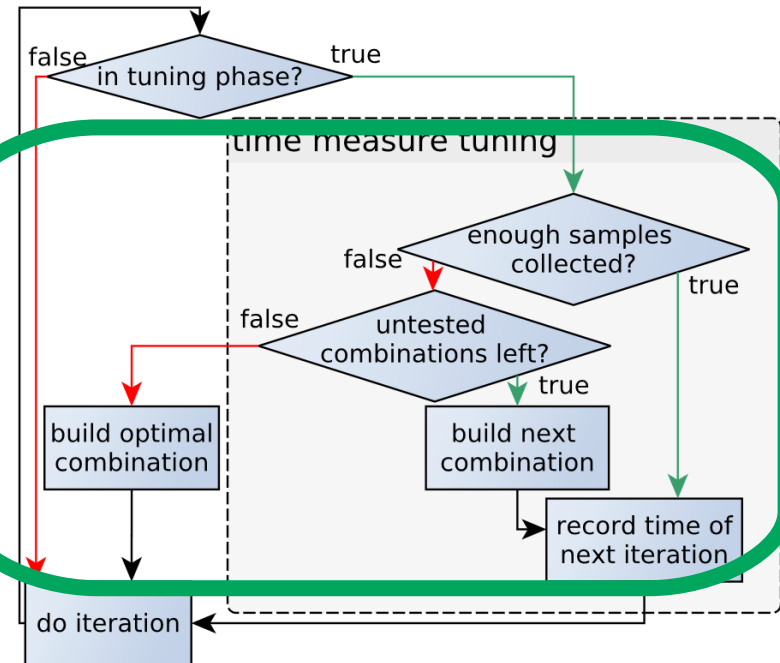
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AutoPas¹

- Improve on “Brute-Force” testing due to outgrowing number of combinations
 - not a nice add-on, but a necessity!
 - work in progress: ML, Bayesian methods



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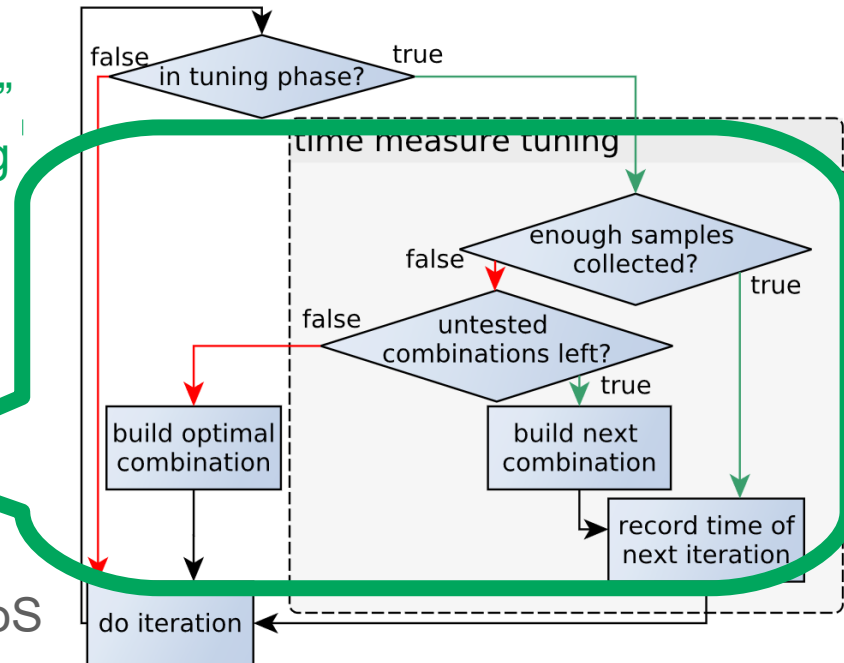
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AutoPas¹

- Improve on “Brute-Force” testing due to outgrowing number of combinations
 - not a nice add-on, but a necessity!
 - work in progress: ML, Bayesian methods
- CUDA & Kokkos support



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AutoPas: Machine Learning for Auto-Tuning

Outline

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- Based on classification via neural network
- Test only five configurations that are deemed best
- Finds (equivalent to) optimal configuration with probability of 99%
- Relies on training data
 - currently, it takes days to train...
 - hardware-dependent training
- Results from a recent evaluation:
 - Full search: **duration: 22.79s # iterations: 27, 0.84 s/it.**
 - ML-based: **duration: 2.05s # iterations: 5, 0.41s/it.**
 - optimal configuration: 0.36s/it (detected by both approaches)
- This is work in progress :-)

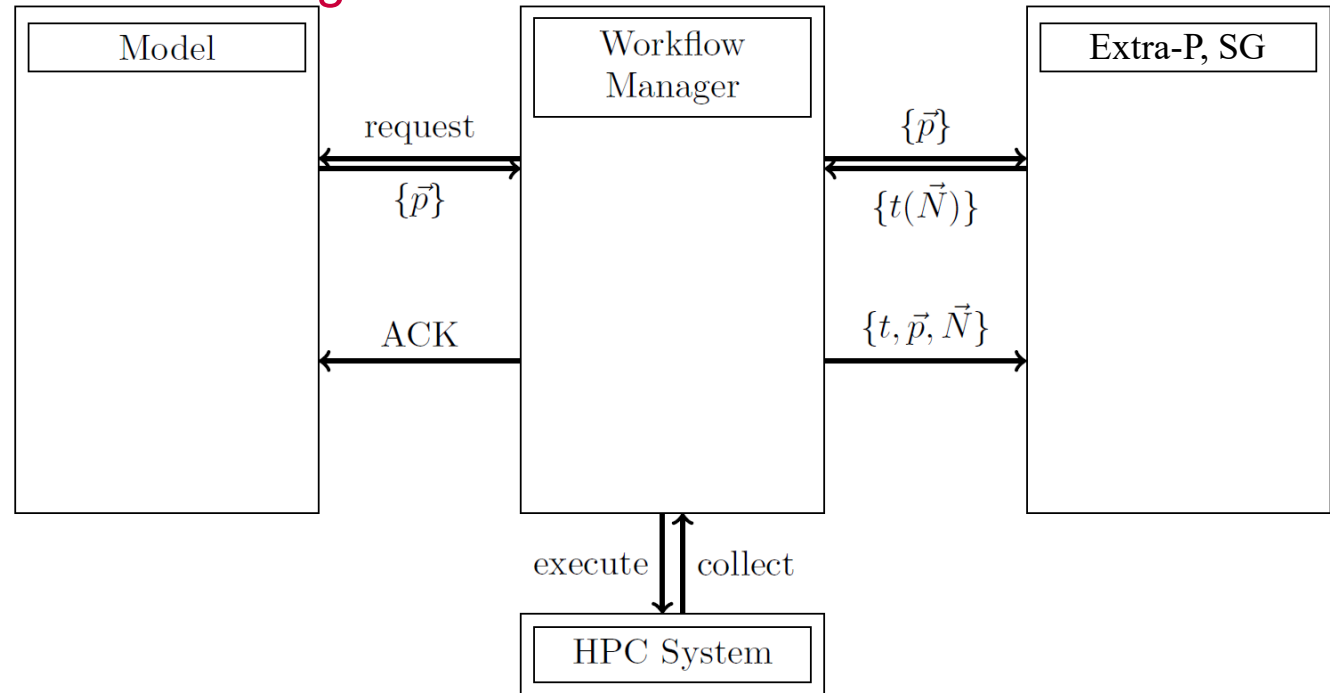
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Workflow Manager



- Last time: Equation of state fitting and vapor-liquid-equilibrium envelope
 - Current work: Sparse Grid-based Perf. Prediction, Adsorption processes³
 - study of adsorption in dispersive systems
 - relevant parameters: temperature, pressure, solid density, interaction fluid-solid
- 350 simulations

³ Heier et al. J. Chem. Eng. Data 64:386-394, 2019

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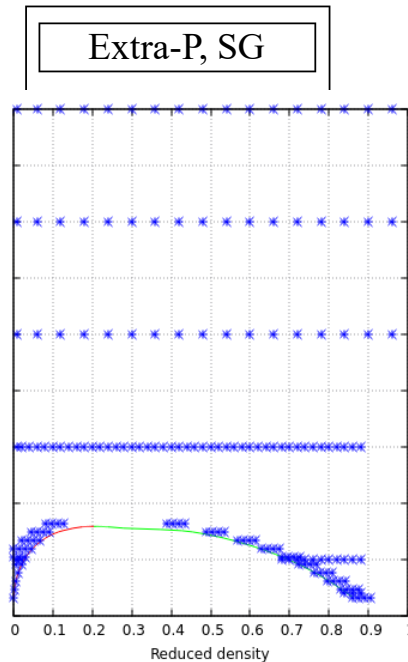
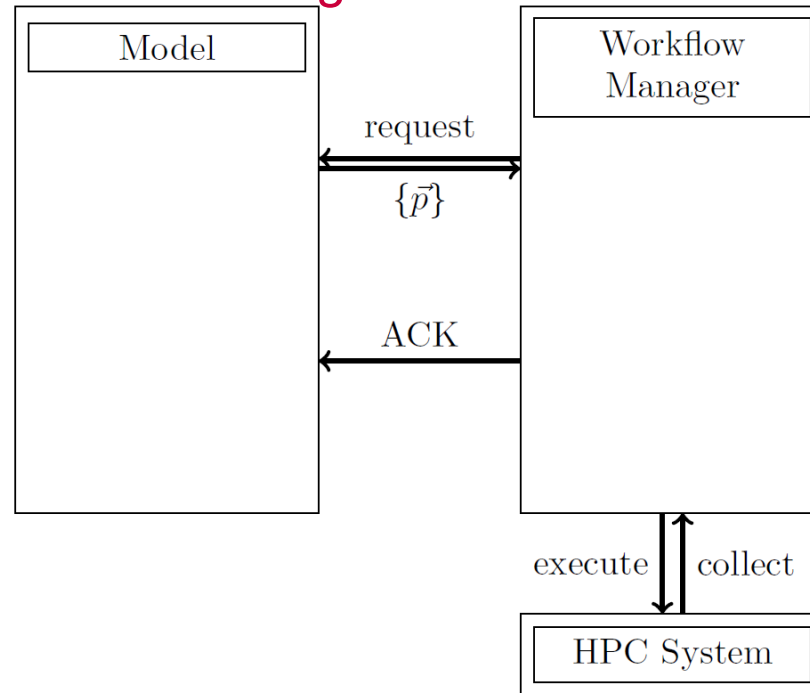


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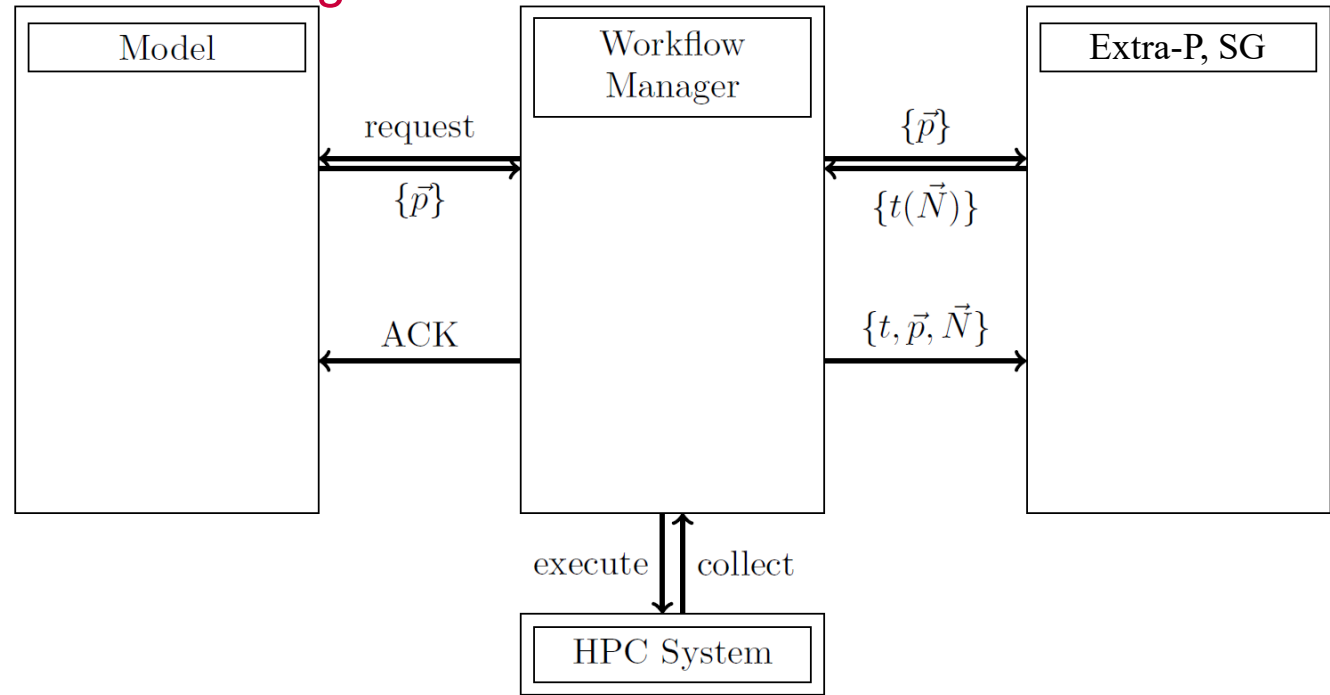
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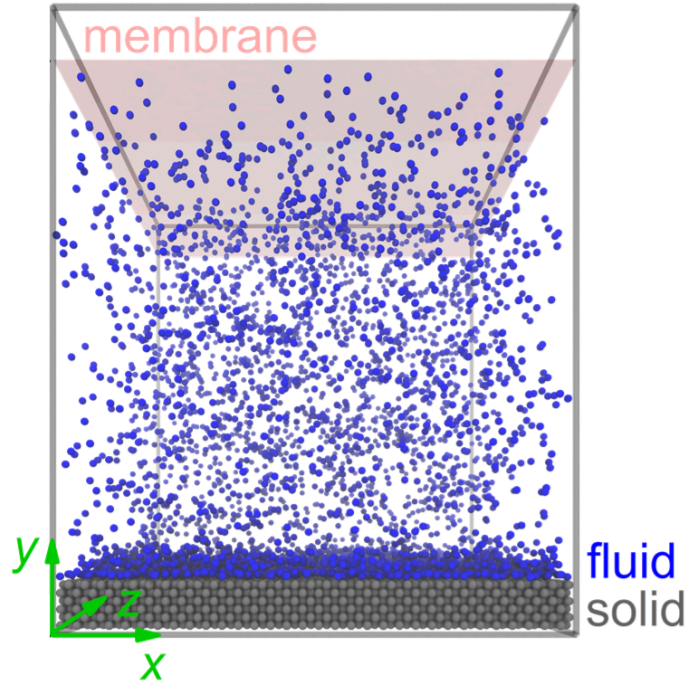
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Workflow Manager and Adsorption



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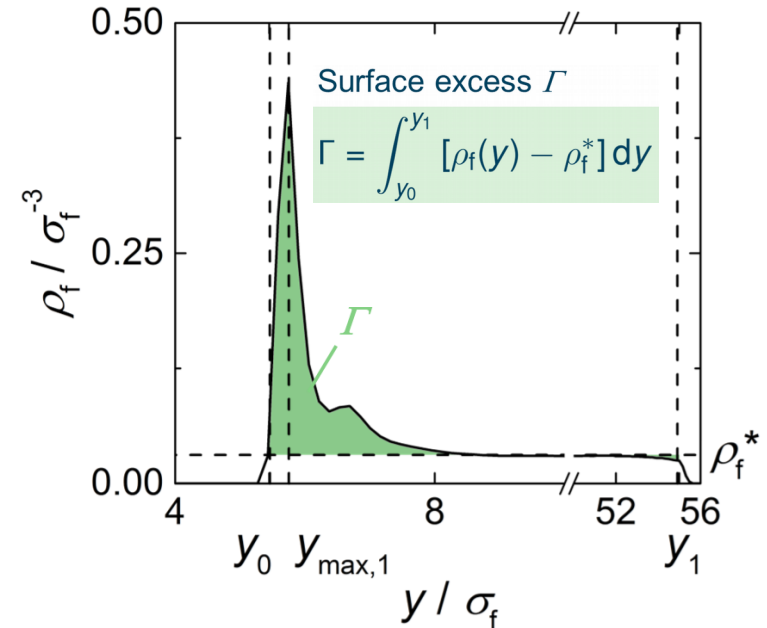
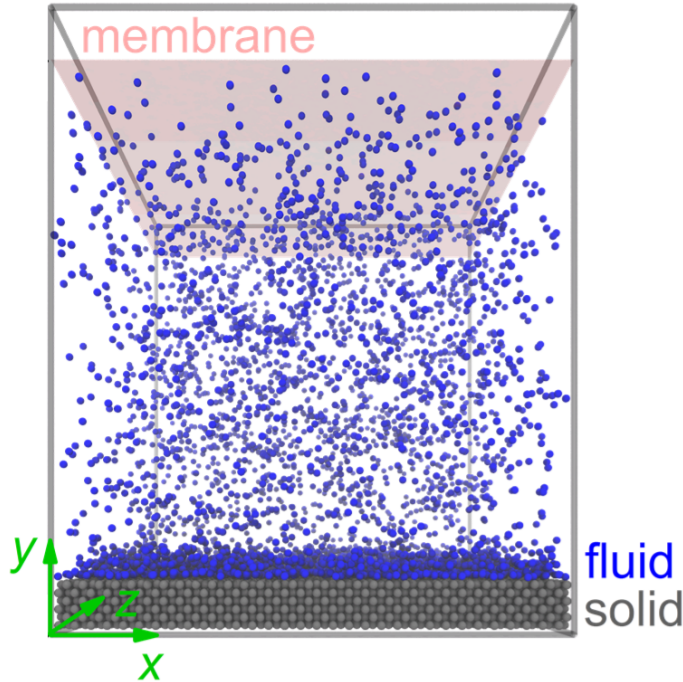


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Workflow Manager and Adsorption

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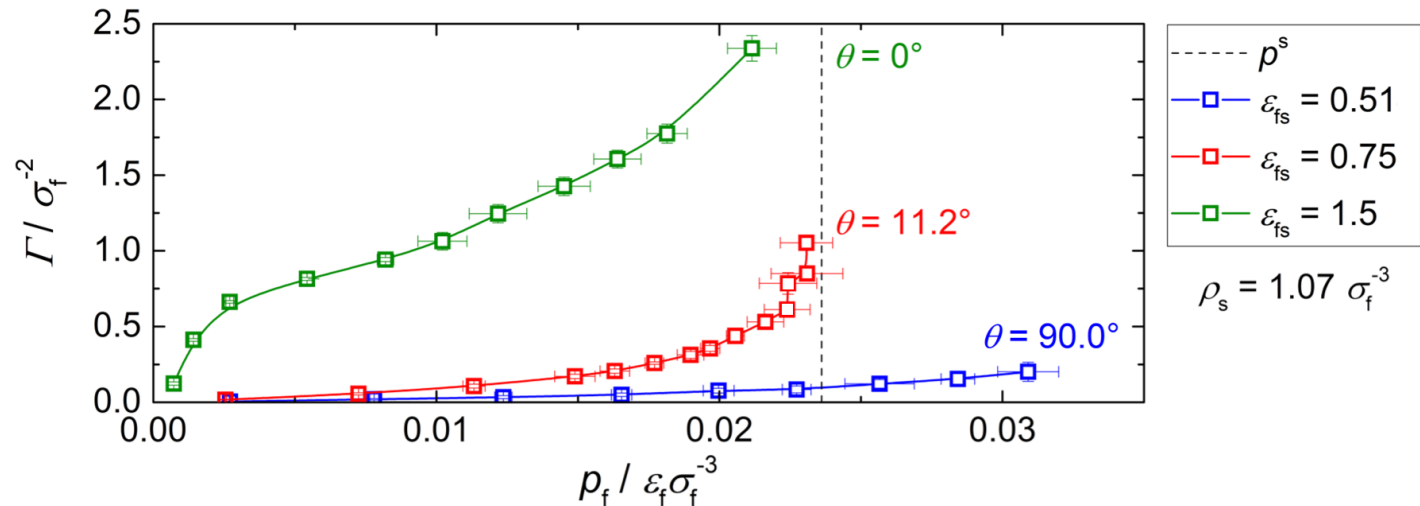
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Workflow Manager and Adsorption

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Influence of fluid-solid interaction



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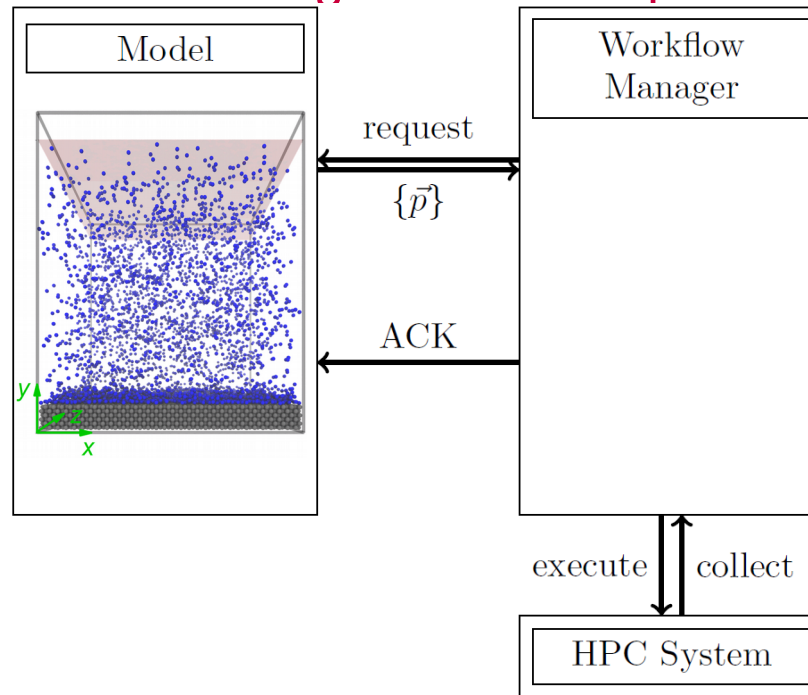
3 Heier et al. J. Chem. Eng. Data 64:386-394, 2019

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Workflow Manager and Adsorption



- Still some user interaction required → work in progress
- Debugging: some SLURM scalability issues...

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Performance Prediction with Extra-P: Higher-Dimensional Parameter Spaces

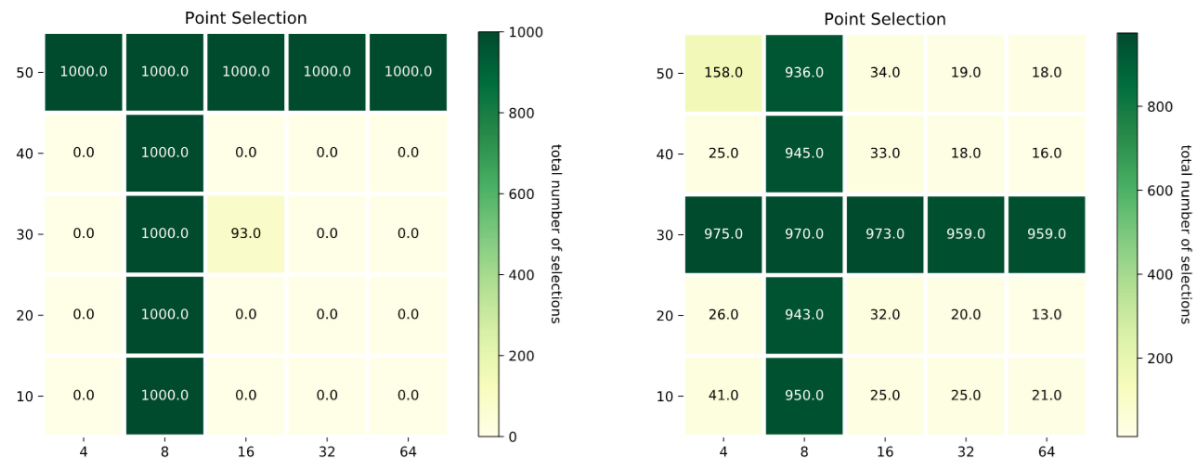
Outline

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- Extra-P
 - Originally:
N parameters $\rightarrow 5^N$ data points to generate performance model normal form

$$f(r_1, r_2, \dots, r_q) = \sum_{k=1}^n c_k \cdot \prod_{l=1}^q r_l^{i_{kl}} \cdot \log^{j_{kl}}(r_l)$$

- Idea: Select subset of data points without degrading accuracy
 \rightarrow reinforcement learning approach



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Performance Prediction with Extra-P: Higher-Dimensional Parameter Spaces

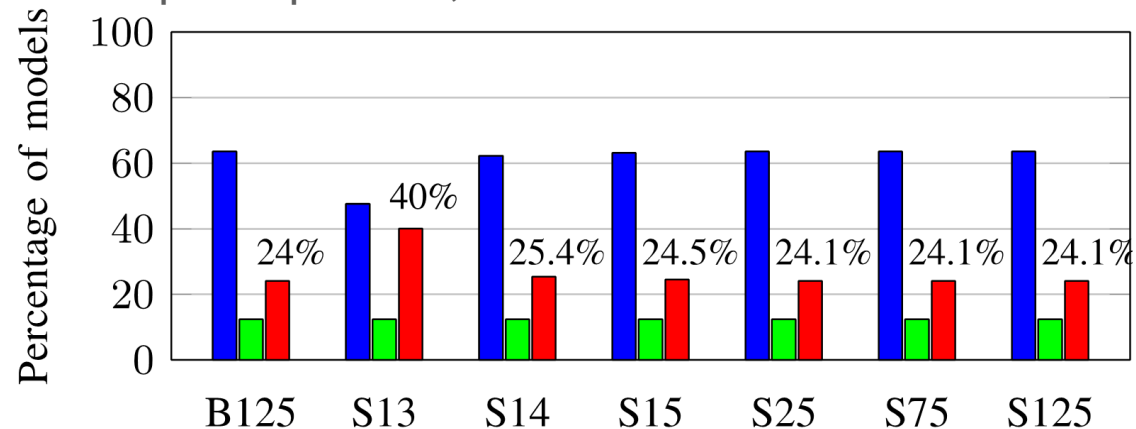
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- Idea: Select subset of data points without degrading accuracy
 \rightarrow reinforcement learning approach
 \rightarrow example: 3 params, 5% noise



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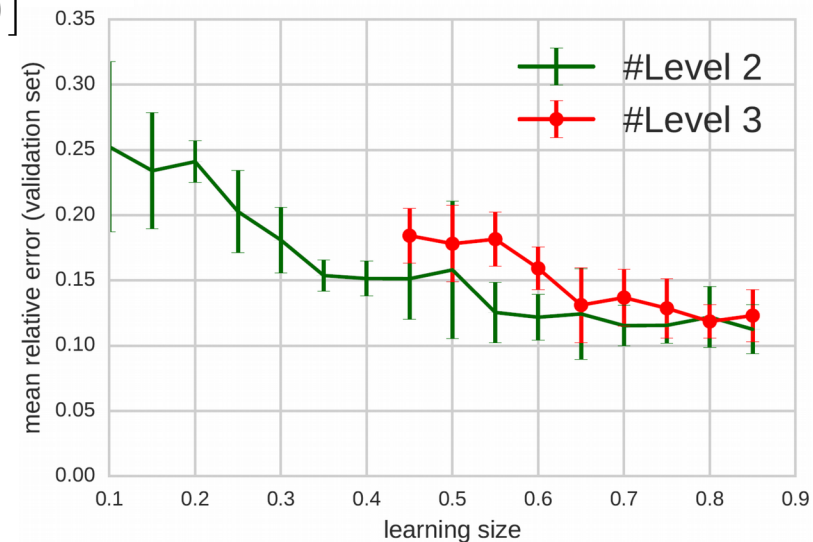
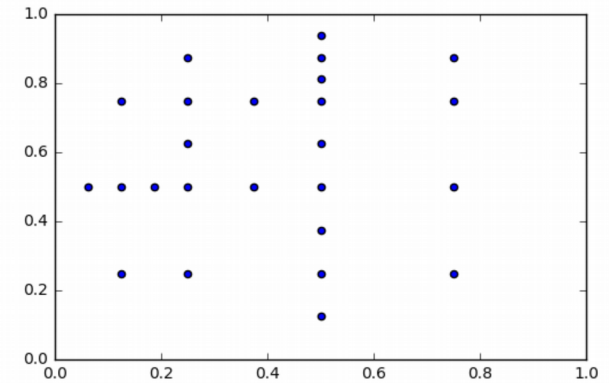


Performance Prediction with Sparse Grids⁴

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- SG=Efficient hierarchical discretization for high-dimensional problems
- Method: Sparse grid regression with local mesh refinement
- Example: Molecular dynamics
 - Particle density $\rho \in [0.3; 0.9]$
 - # particles $N \in [10^3; 10^5]$
 - Cut-off radius $r_c \in [1.2; 4.5]$
 - Blocksize $b \in [10; 10^3]$
 - # MPI procs $p \in \{1, 2, 4, 8\}$



4 P. Neumann. Sparse Grid Regression for Performance Prediction Using High-Dimensional Run Time Data. Euro-Par Workshops (PMACS) 2019. Accepted

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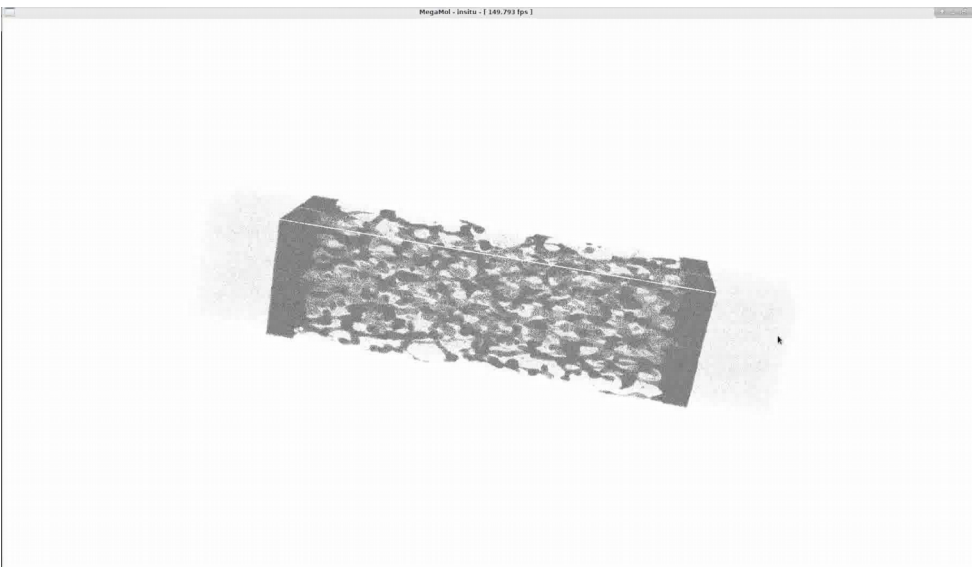
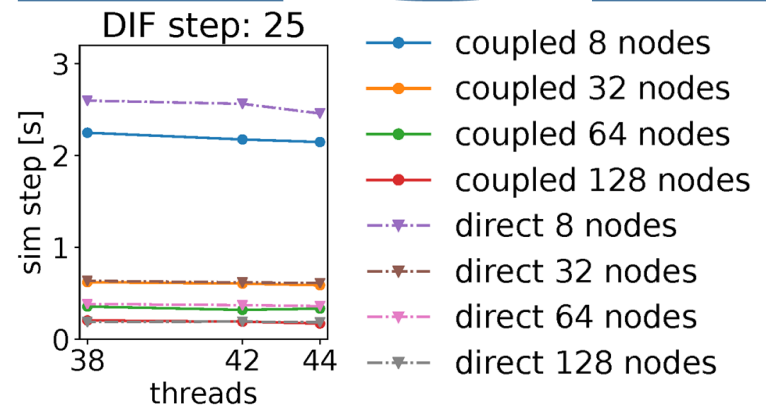
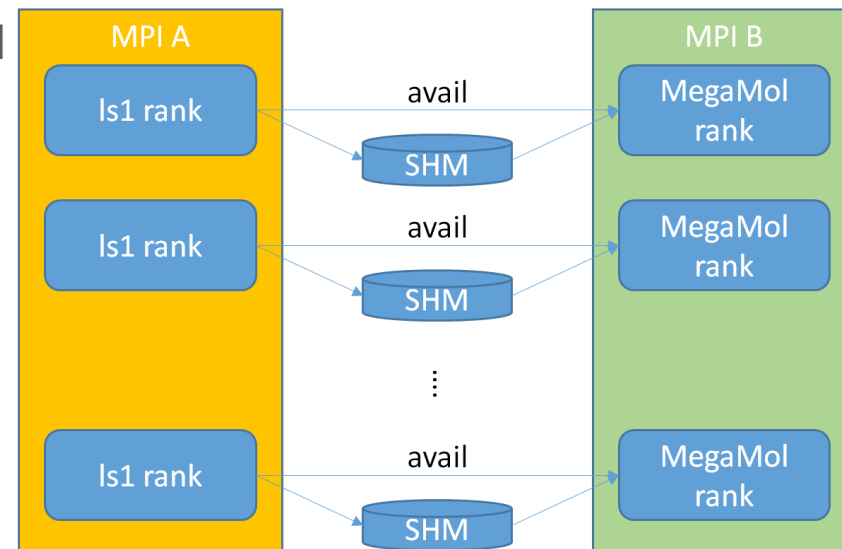
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In-Situ Visualization

Outline

- AutoPas
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- **In-Situ Visualization**

- MegaMol: Integration in Is1
 - Data exchange via shared memory
 - Different MPI worlds
 - if visualization fails, simulation continues

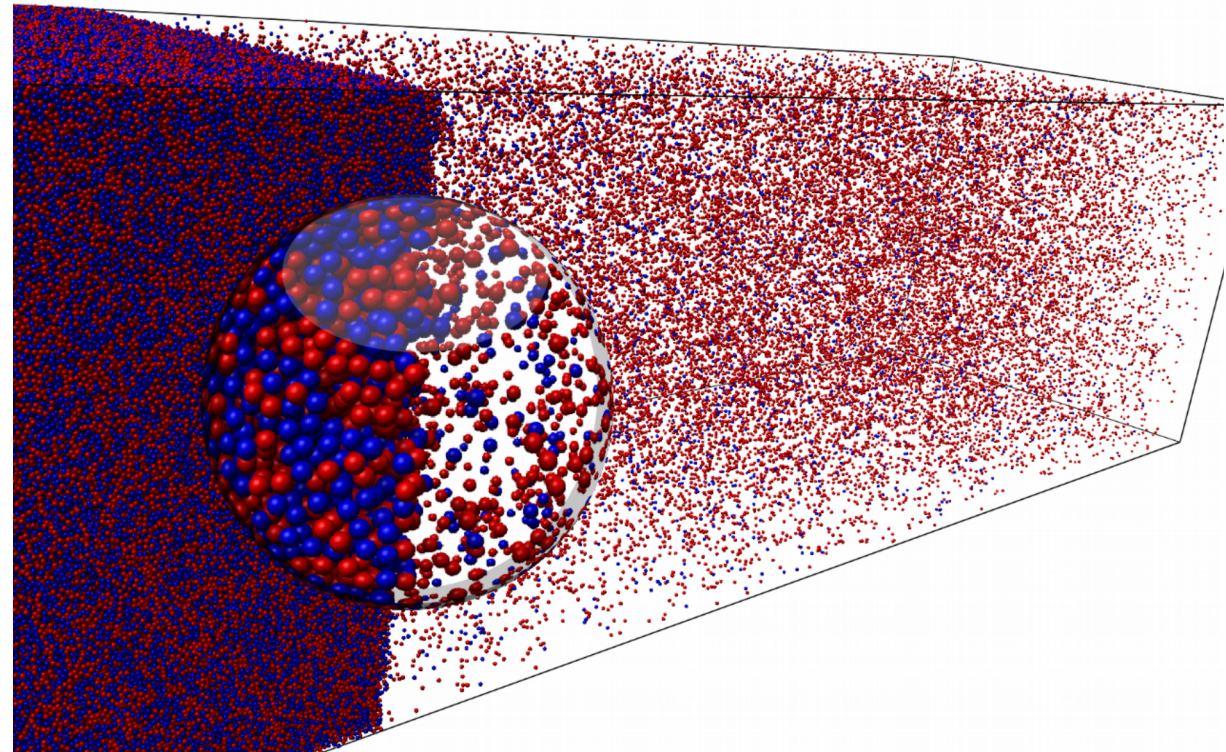


Resilience

Outline

- AutoPas
- Workflow Manager
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- **In-Situ Visualization**

- Checkpointing-based
- Future work: evaluation of quality for molecular films



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Outline

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Summary

- AutoPas
 - Integration in Is1, first application-relevant tests
 - Work in progress: improved tuning procedure, CUDA support
- Workflow Manager
 - New scenario: adsorption processes
→ first (hand-steered) workflow solution using Is1
 - Work in progress: Scheduling with performance prediction, integration of components, scalability
- Performance Prediction
 - New methods for high-dimensional parameter spaces (Extra-P, Sparse Grids)
 - Work in progress: Integration with Workflow Manager
- In-Situ Visualization: Integration of MegaMol and Is1

P. Neumann and colleagues acknowledge funding by the Federal Ministry of Education and Research, grants 01IH16008, project TaLPas.

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- In-Situ Visualization: **Integration of MegaMol and Is1**

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- ISC HPC 2020: Project Posters, Submission Deadline: 12 Feb 2020
→ <https://www.isc-hpc.com/project-posters-2020.html>
- (If you know someone who's) Interested in a PhD/Postdoc on HPC, multiscale flow simulation, molecular dynamics, performance prediction, etc.
→ contact me, philipp.neumann@hsu-hh.de :-)

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