

# MDProcessing.jl: Julia Language Application to the Molecular Simulation Trajectory Analysis

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# Molecular Dynamics Simulations

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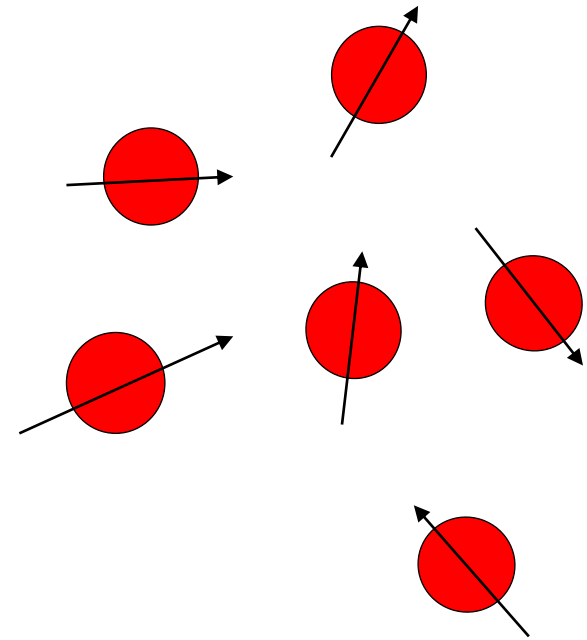
$$\mathbf{F}_i = m_i \mathbf{a}_i$$

Forces are computed from an analytical potential (force field)

$$\mathbf{F} = -\nabla V(r) \quad \mathbf{F}_i = \sum_{j \neq i} \mathbf{F}_{ij}$$

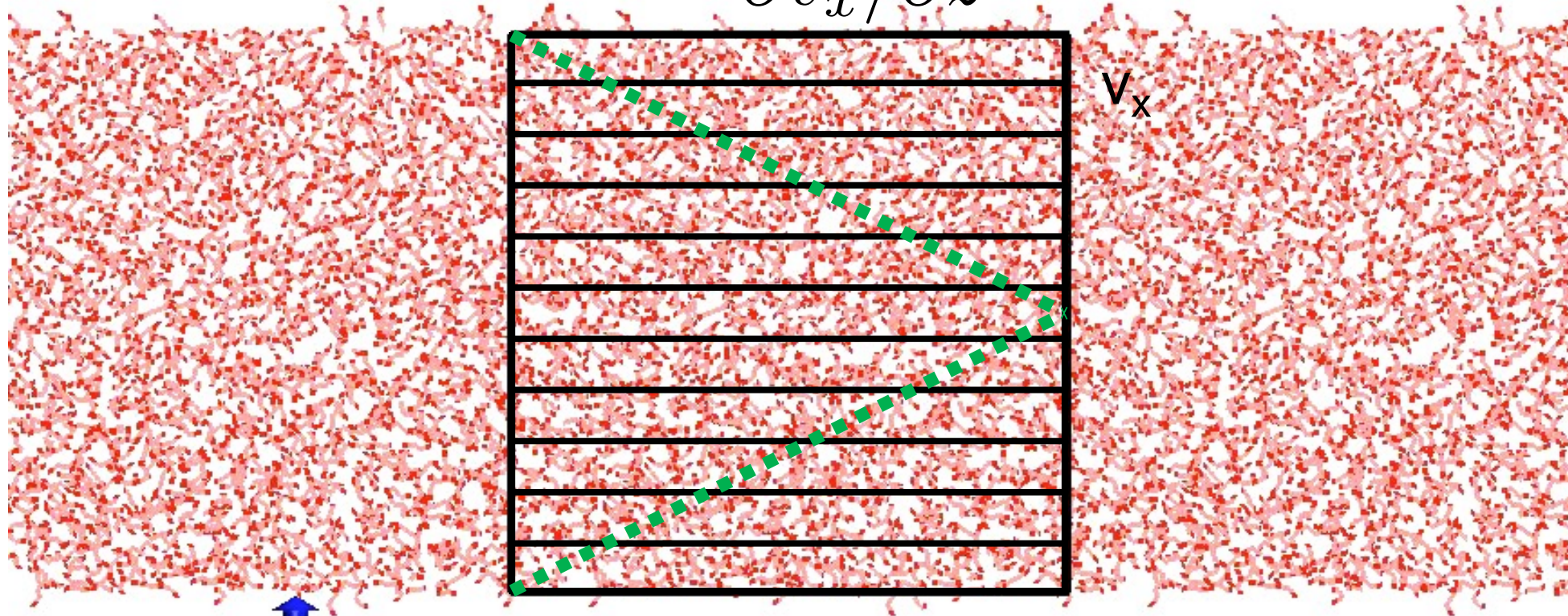
Equations of motion are solved using a finite-difference algorithm

$$\mathbf{r}_i(t + \Delta t) = \mathbf{r}_i(t) + \mathbf{v}_i(t)\Delta t + \mathbf{a}_i(t) \frac{\Delta t^2}{2}$$
$$\mathbf{v}_i(t + \Delta t) = \mathbf{v}_i(t) + \frac{\mathbf{a}_i(t) + \mathbf{a}_i(t + \Delta t)}{2} \Delta t$$



# Example: shear flow

$$\eta = \frac{j_z(p_x)}{\partial v_x / \partial z}$$



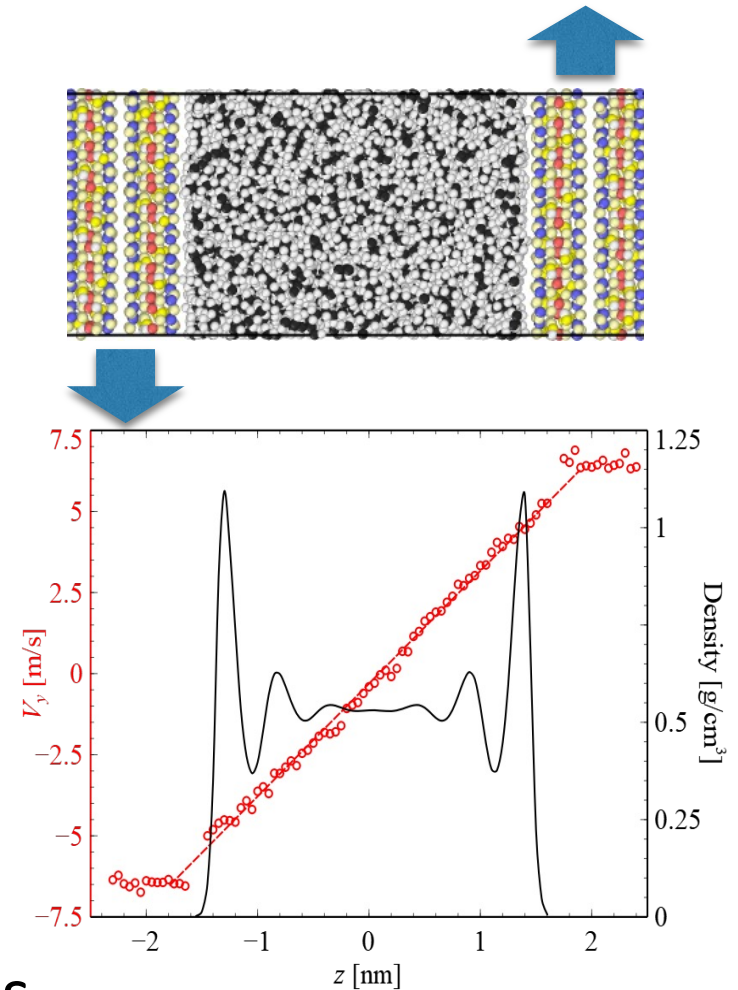
$T = 330 \text{ K}; L \sim 6 \text{ nm}; N = 80\text{k};$

**(Muller-Plathe)** Muller-Plathe, Phys Rev E, 59, 4894-4898 (1999).

*Подробности:* Кондратюк Н Д, Писарев В В  
"Теоретические и вычислительные подходы к предсказанию вязкости жидкостей"  
УФН, 2023

# Molecular Dynamics Simulations

- Computation of thermophysical properties involves
  - Structural analysis of instantaneous states
  - Averaging properties over a trajectory
- Typical system: 1k – 10M particles
- Typical simulation time: 10 ps – 100 ns (10k – 100M time steps)
- Large amount of data per trajectory, analysis becomes computationally challenging task



# Tools for Simulation

- Primary requirements:
  - Fast implementation of standard integration methods, thermostats, force fields
  - Correctness and robustness
- Simulation packages: Gromacs, LAMMPS, DL\_POLY, HOOMD
- Highly optimized for CPU, GPU or both
- Almost necessary to use a programming language with a low-level control: C++ or Fortran (maybe Rust, we may see new packages released in the next few years)

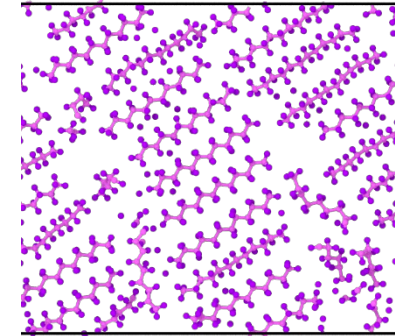
# Trajectory analysis

- The most time-consuming part is already done
  - Less pressure on performance
- Two kinds of needs
  - Standardized analysis
  - Exploratory analysis, development of new structural and dynamic properties

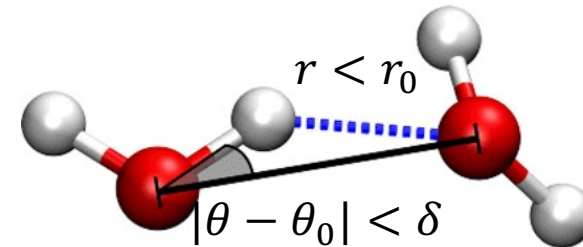
# Non-standard properties

- Correlation between interatomic separation and velocity directions

- Characterize the degree of molecular ordering



- Identification of H-bonds



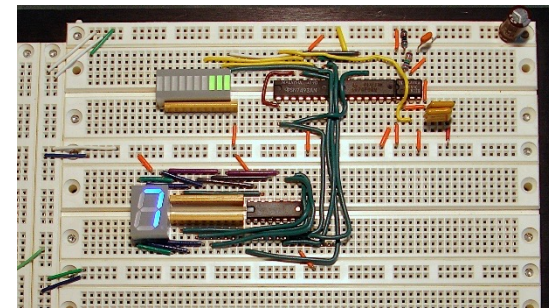
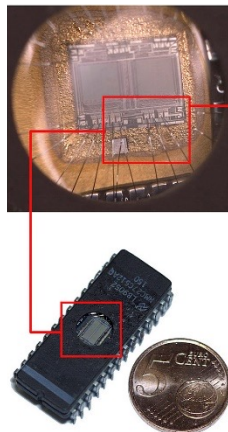
# Existing approaches

Static implementation as an executable program (TraVIS, LAMMPS, Gromacs)	Scripting language binding (VMD, OVITO)
✓ Fast	✓ Fast “standard” methods implemented in C++
✓ Can be done on the fly within MD simulation software (LAMMPS, Gromacs)	✓ Easy user extension
	✓ Interop with programs written in the scripting language – processing is easily incorporated into a higher-level pipeline
• Not extensible without rebuilding the executable	
• Interop with other programs only via file output	• User-defined extensions will be less performant than “built-in” operations



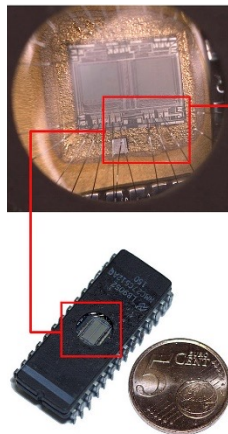
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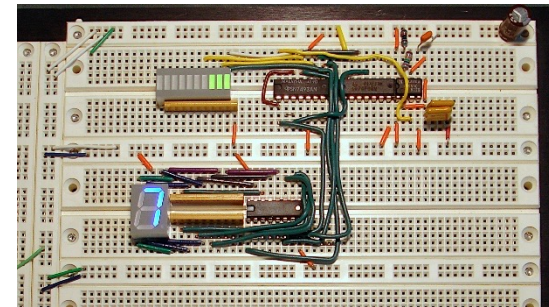


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Middle ground?



# Why use Julia programming language?

- Dynamic typing and duck typing – good for exploratory programming and data analysis
- Type inference: in a carefully-written program, types can be statically inferred
- JIT compilation – carefully written code can be statically type-inferred and compiled to fast native code
  - Core computations can be implemented in pure Julia
  - User-defined data types are as performant as built-in types
  - Raw loops are fast - no need to jump through hoops to offload heavy computations to a specialized numerical library
- Convenient out-of-the-box data containers such as arrays, tuples, dictionaries, sets, multiple useful data structures provided by third-party packages

# Why use Julia programming language?

- Parallel features:
  - Native task-based multithreading
  - Distributed-memory computing via built-in Distributed.jl or MPI bindings
  - CUDA programming via CUDA.jl

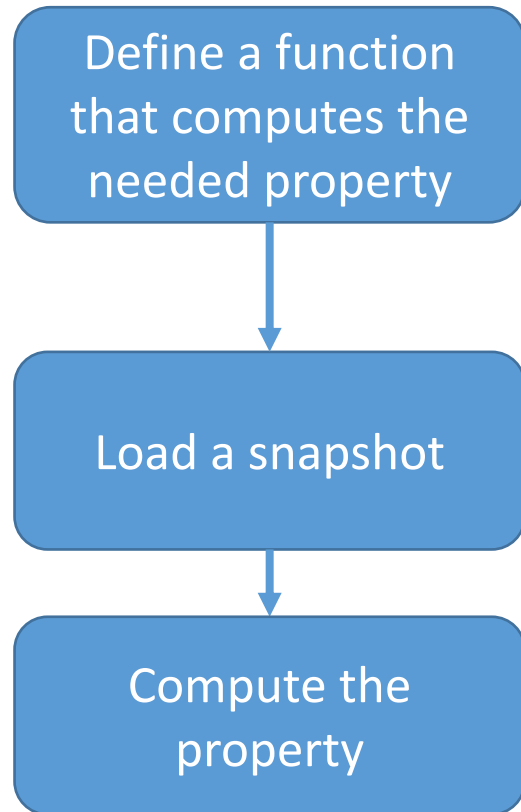
# MDProcessing.jl package

- A package for computing properties from MD trajectories
- Written in Julia
- Data format: LAMMPS dump files (XYZ and LAMMPS data files are planned)
- Built-in analysis functions are defined, user extensions in Julia are possible

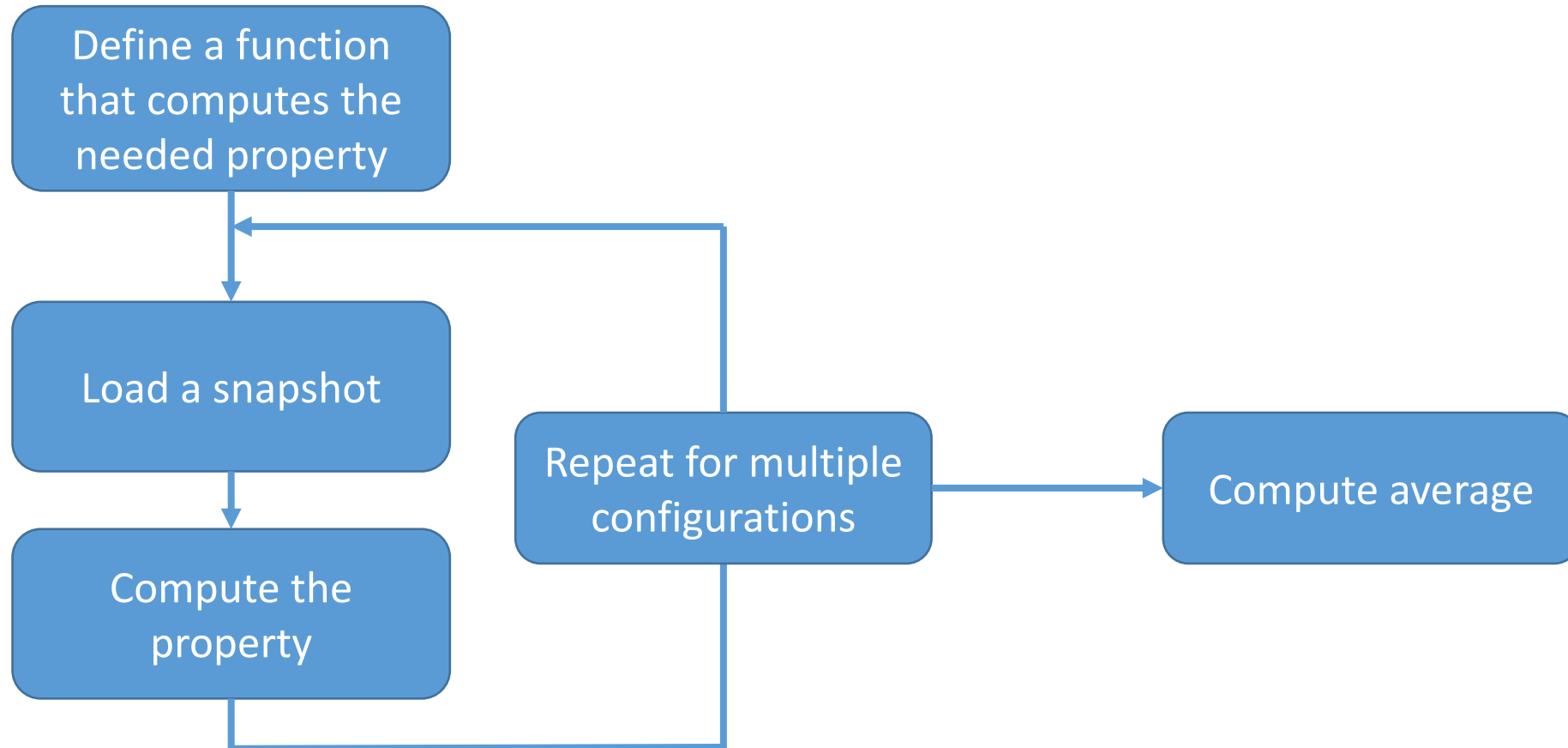
# MDProcessing.jl intended workflow

Define a function  
that computes the  
needed property

# MDProcessing.jl intended workflow



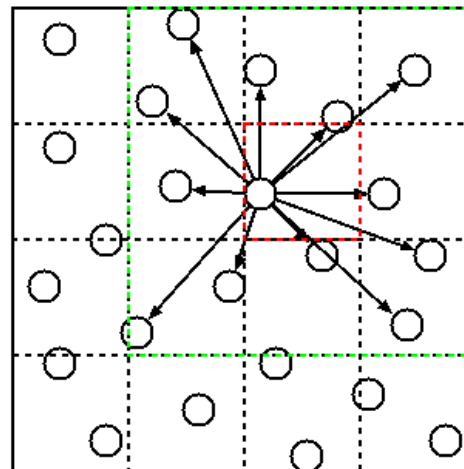
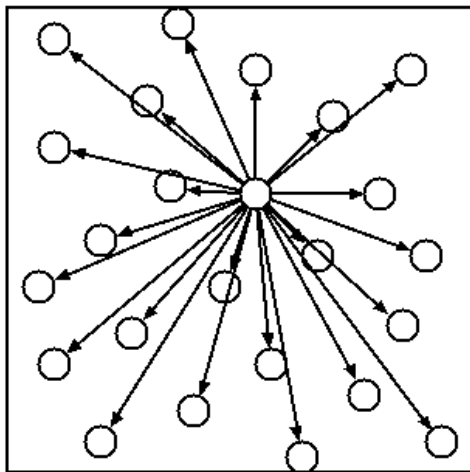
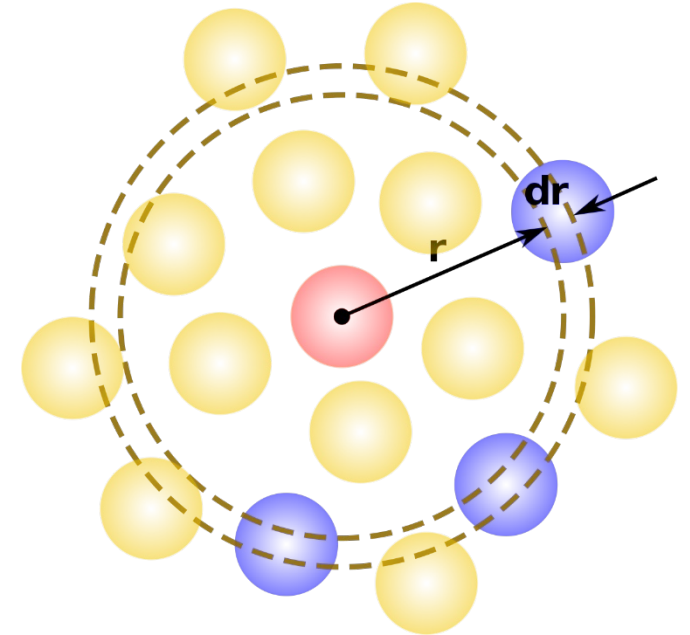
# MDProcessing.jl intended workflow





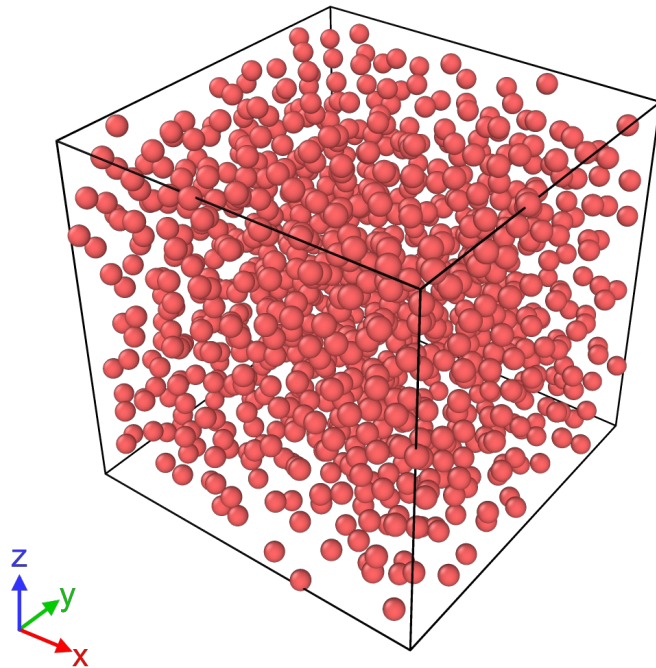
# Performance assessment

- Test problem: computing the radial distribution function
- Benchmarks one of the costly operations in particle simulation analysis – neighbor search
- Algorithmic complexity  $O(N^2)$  with naïve search,  $O(N \times R_c^3)$  with cell list structure

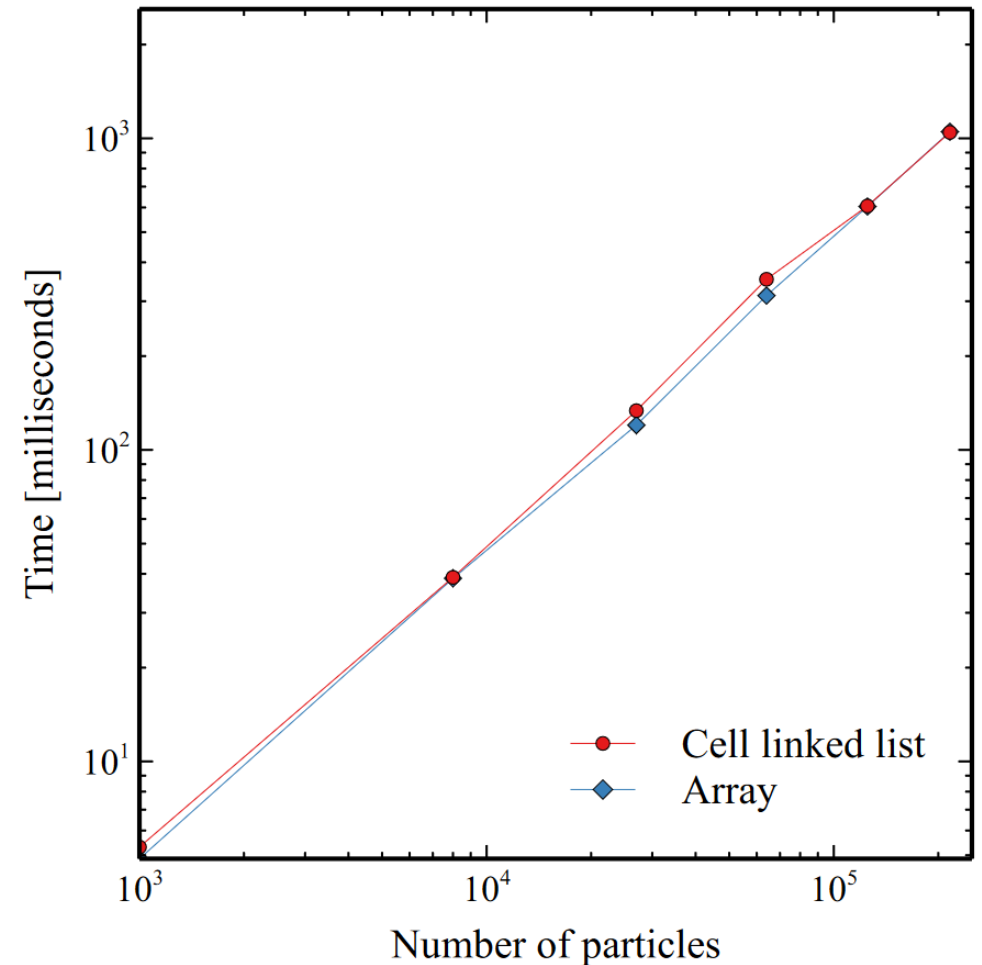


# Performance assessment

- Test system: Lennard-Jones fluid
- Constant density  $\rho = 0.75$
- Varying system size and cut-off radius

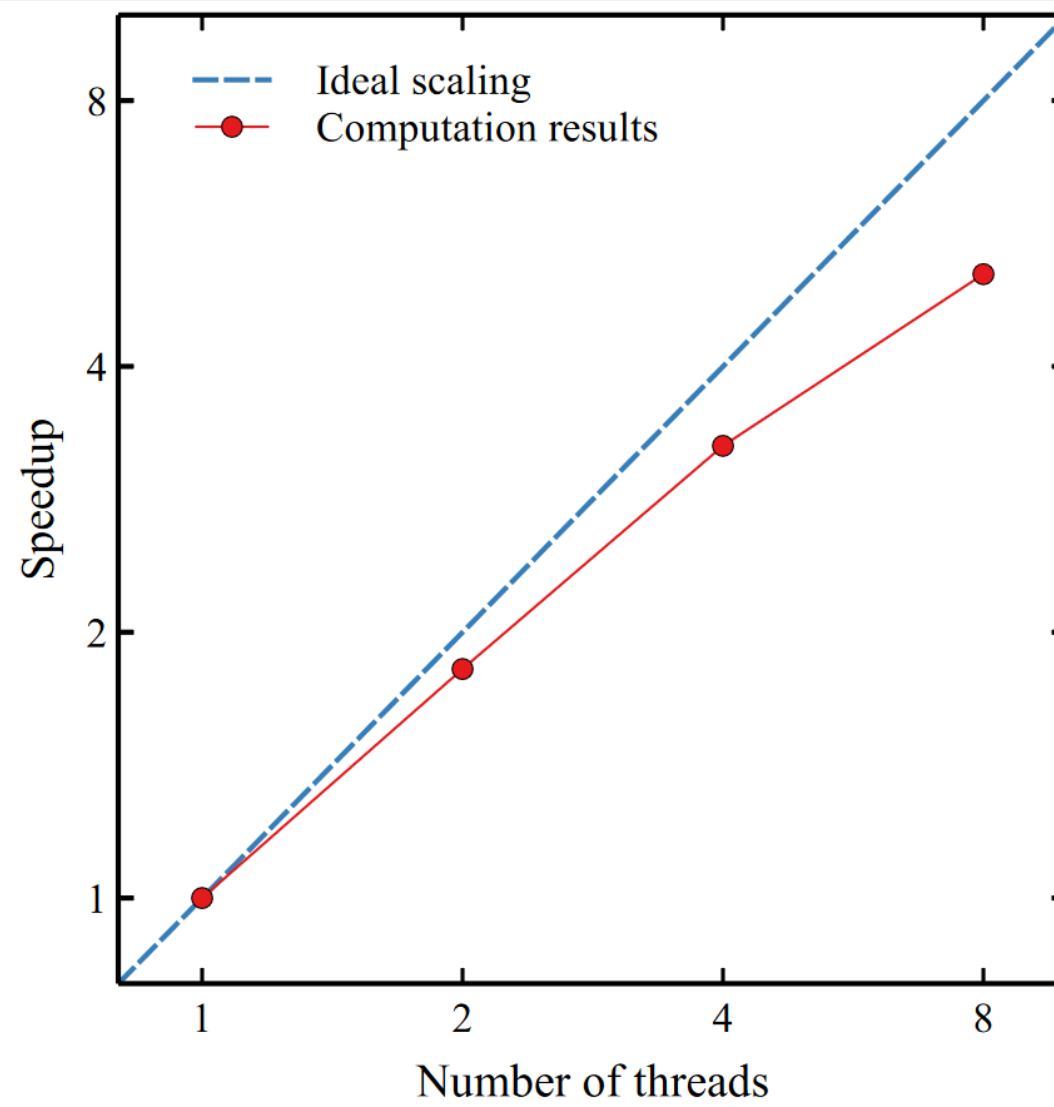
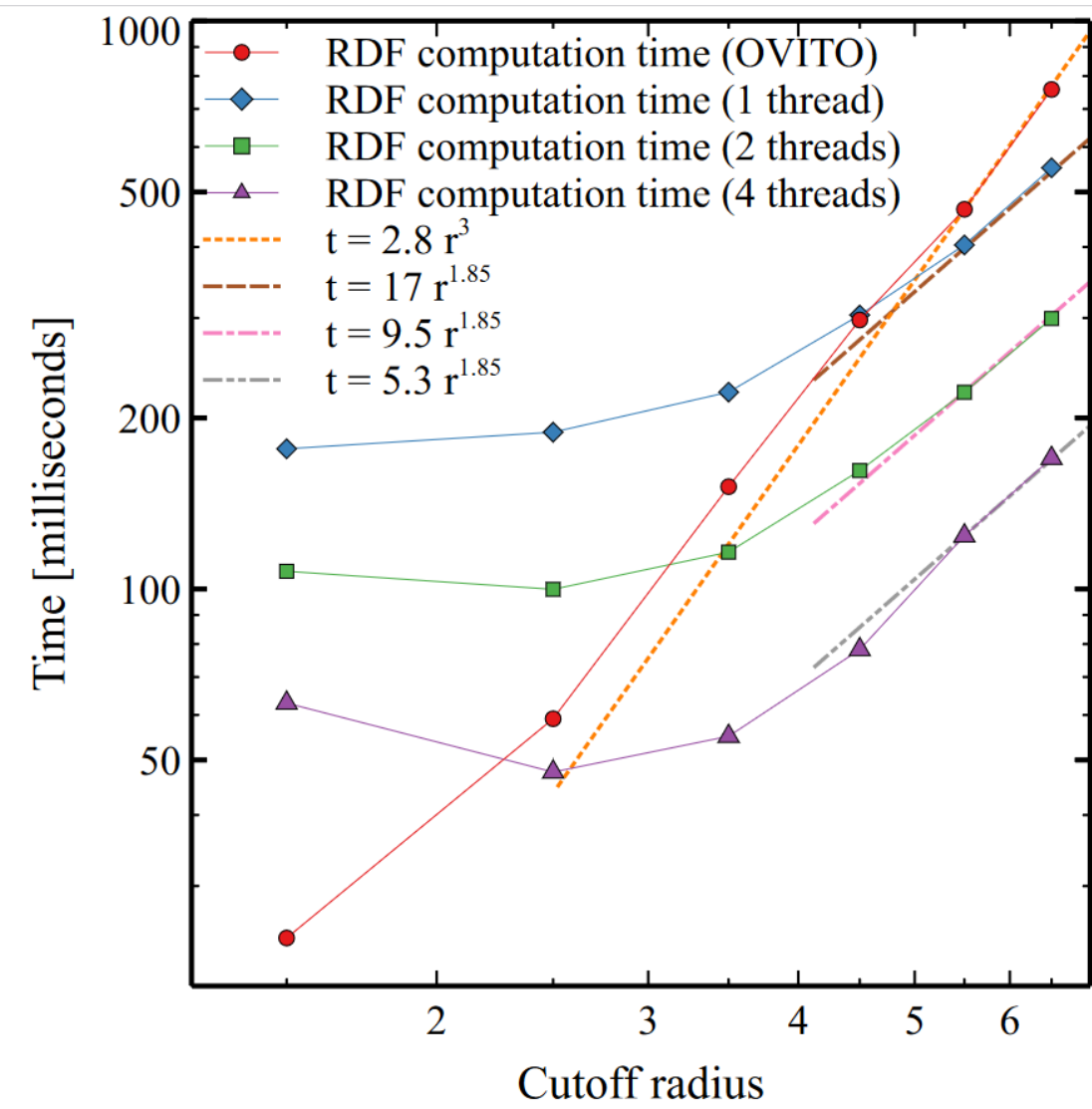


RDF compute time for different system sizes ( $R_c = 2.5$ )

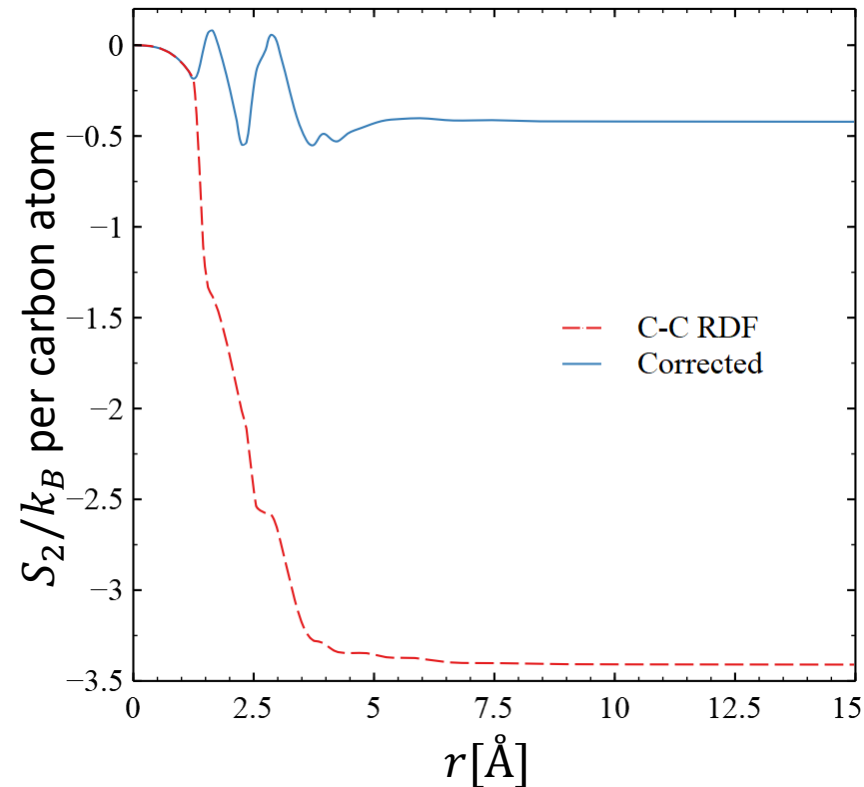
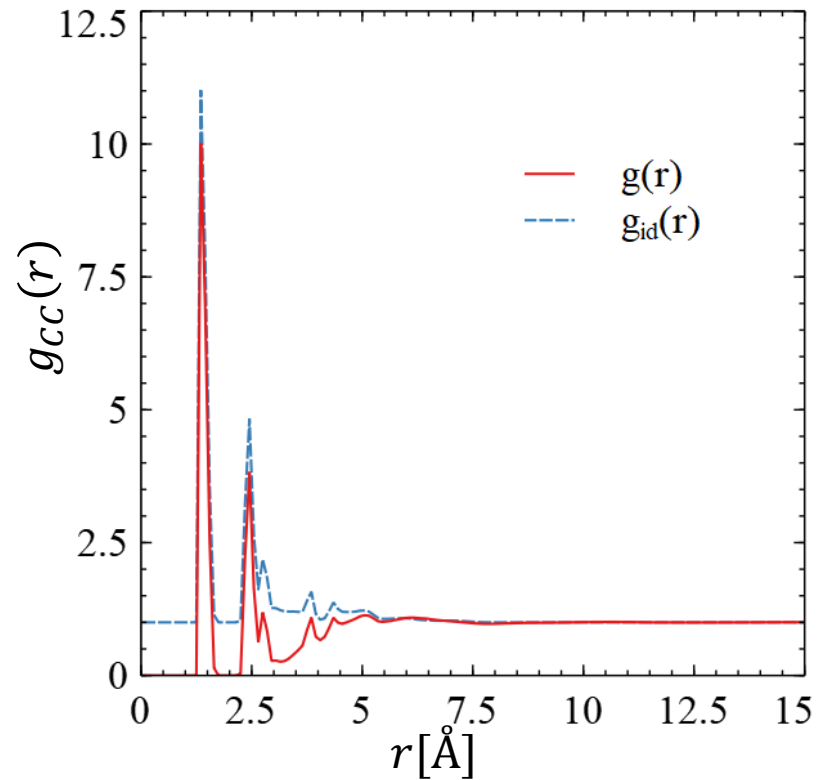


# Performance assessment

## Effect of cut-off radius and multi-threading

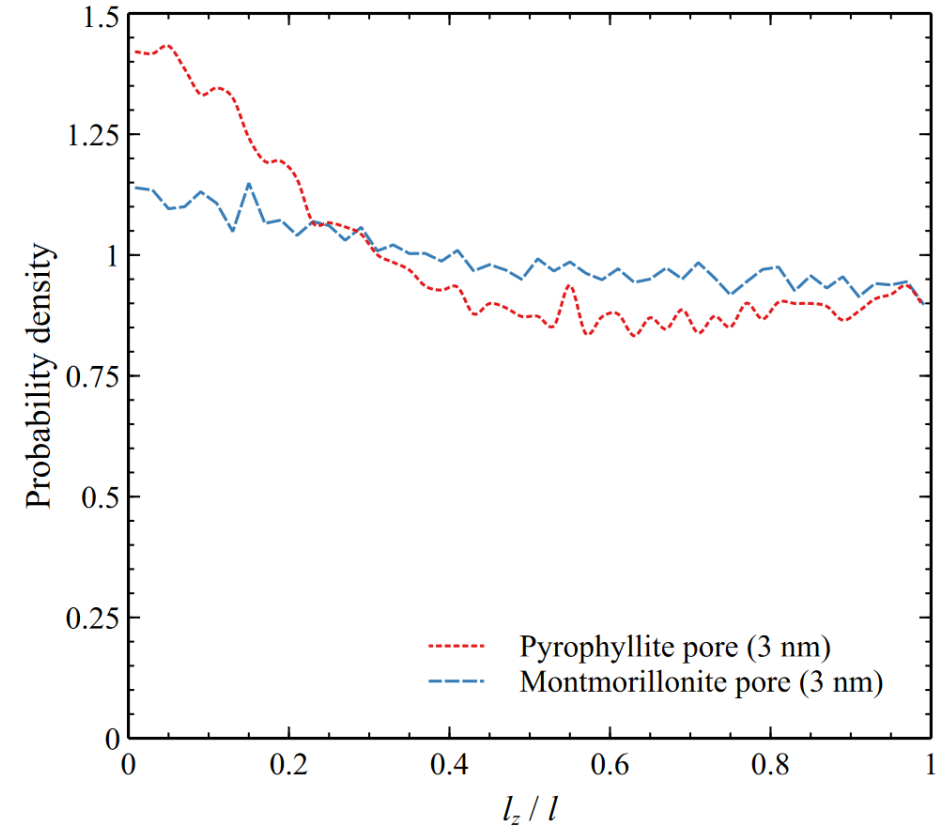
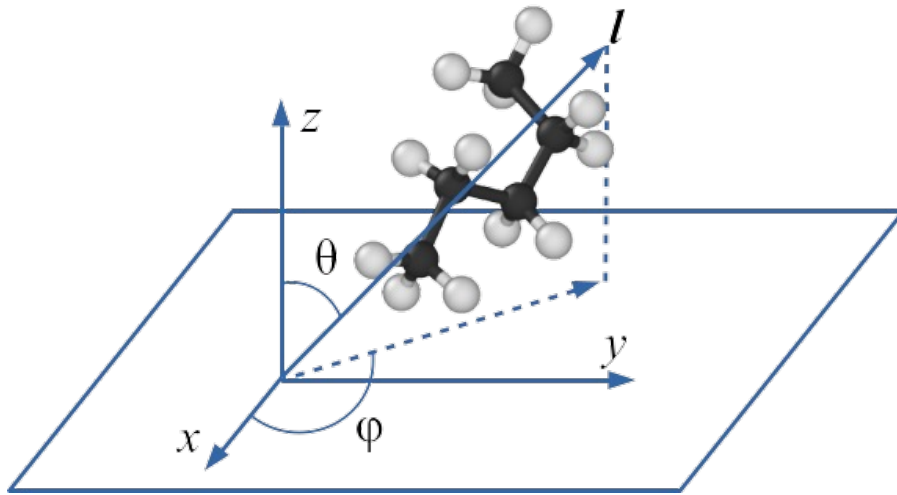


# Some results



Computation of RDFs in molecular systems separated into intra- and intermolecular contributions  
Nikitiuk, Salikova, Kondratyuk, Pisarev // J Mol Liq 2022

# Some results



Straight-line fitting and orientational distributions of molecules  
Pisarev & Kalinichev // J Mol Liq 2022

# Conclusions

- Julia language is suitable for performance-demanding computational tasks
- Dynamic typing and scripting nature of the language make it convenient to do exploratory analysis and prototyping new ideas
- Performance-critical parts of the analysis can be optimized *in the same language*
- The core functions of MDProcessing.jl are shown to have the performance comparable to OVITO (written in C++)
- Project repository: <https://gitlab.com/pisarevvv/mdprocessing.jl/>