



International School
“Computer simulation of advanced materials”
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Molecular dynamics

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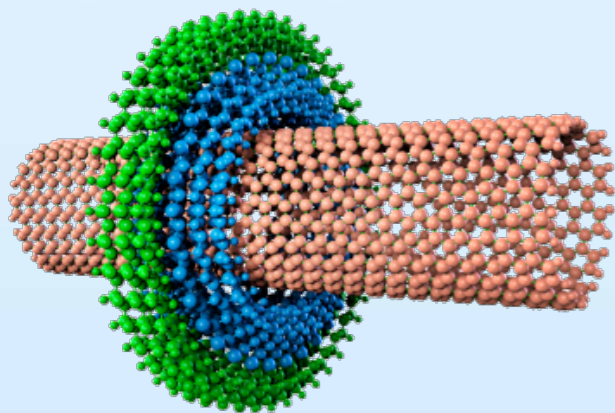
alex@molsim.org

Outline

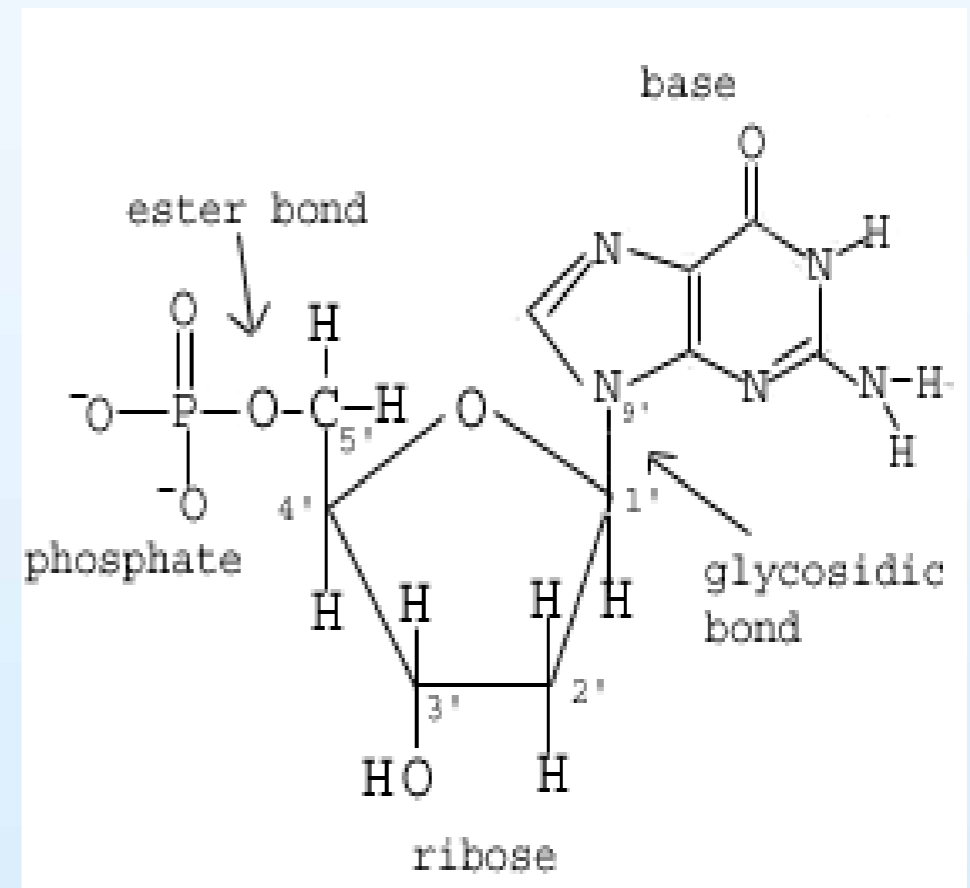
- The idea
- Approximations and validity
- Molecular mechanics, force fields
- Numerical simulations
- Technical details
- High performance computing
- Examples

Motivation

Matter at the nano scale



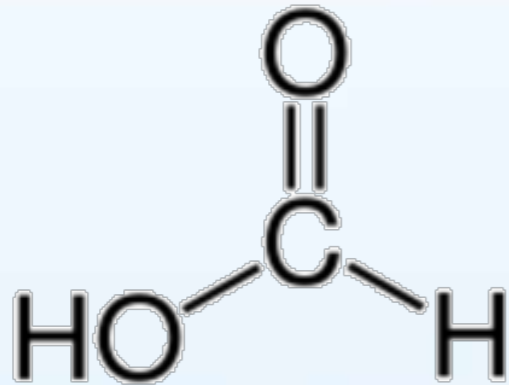
Atom scale $\sim 10^{-10}$ m



Molecules = atoms
connected with bonds

The idea

Molecular dynamics simulations



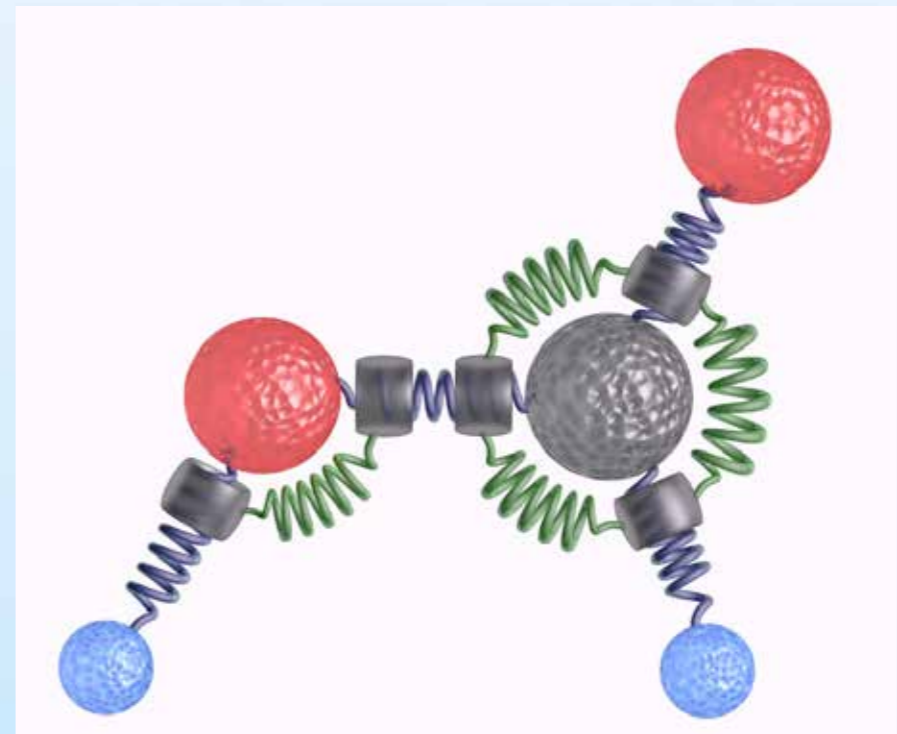
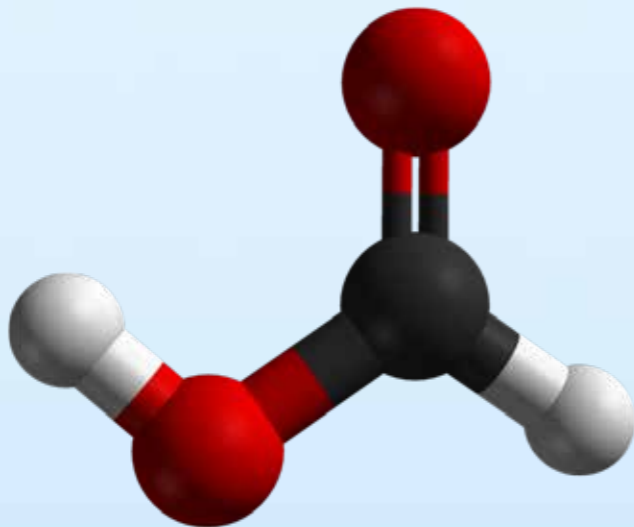
Formic acid

Laws of classical mechanics

$$\vec{F} = m * \vec{a}$$

Principles of statistical physics

$$P \propto \exp(-E/kT)$$



Fundamentals: from quantum to classical

$$i\hbar \frac{\partial}{\partial t} \Psi = \hat{H} \Psi$$

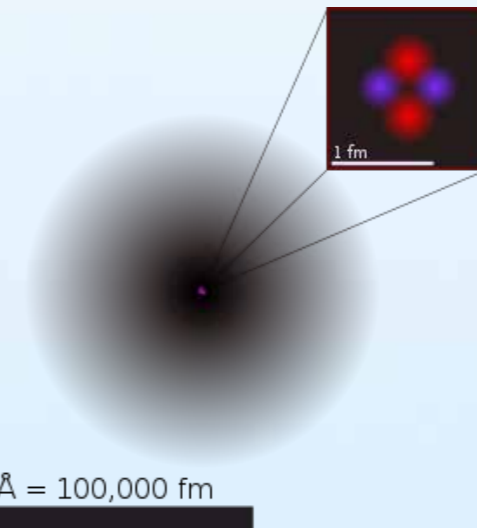
Time dependent non-relativistic
Schrödinger equation

$$E \Psi(\mathbf{r}) = \frac{-\hbar^2}{2m} \nabla^2 \Psi(\mathbf{r}) + V(\mathbf{r}) \Psi(\mathbf{r})$$

Time independent
Schrödinger equation for a single particle

$$\Psi_{\text{total}} = \psi_{\text{electronic}} \times \psi_{\text{nuclear}}$$

Born-Oppenheimer approx.



electron mass approx. 1/1836
that of the proton

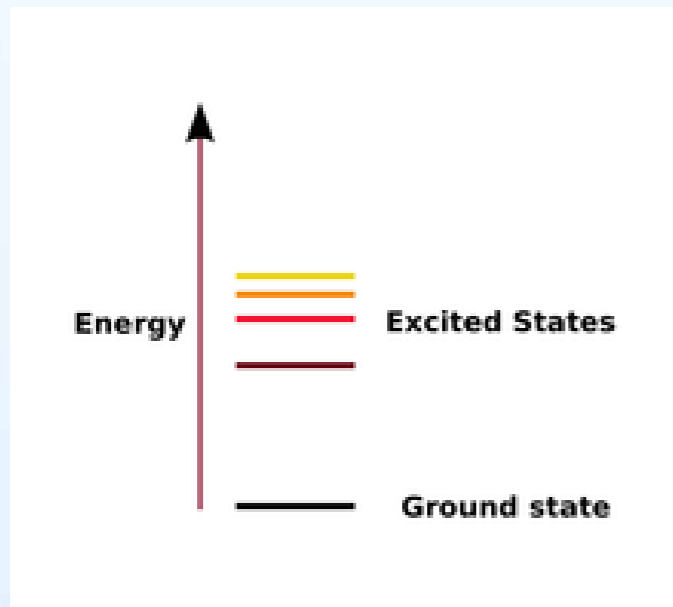
$$H_e(\mathbf{r}, \mathbf{R}) \chi(\mathbf{r}, \mathbf{R}) = E_e \chi(\mathbf{r}, \mathbf{R})$$

$$E_e(\mathbf{R})$$

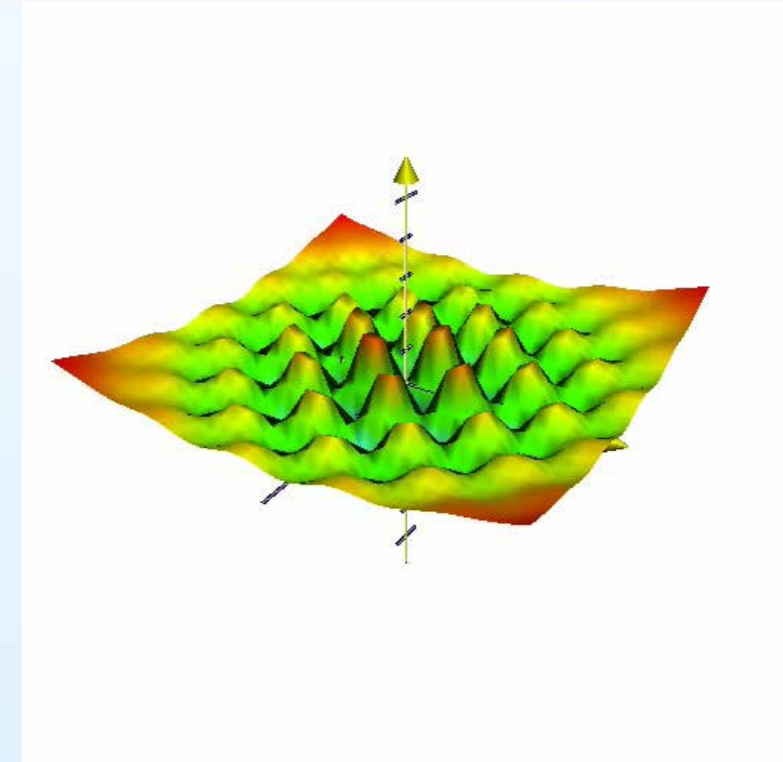
- potential energy
surface

Fundamentals: from quantum to classical

$E_e(\mathbf{R})$ - potential energy surface



Ground state of
electronic wave function



The nuclei movements are considered classical

$$i\hbar \frac{\partial}{\partial t} \Psi = \hat{H} \Psi \quad \longrightarrow \quad \frac{d\mathbf{p}}{dt} = -grad(E_e(\mathbf{R}))$$

Approximations, restrictions and gains

1. BO approx. (adiabatic approx.)

No nonadiabatic processes (charge-transfer reactions, photochemistry)

2. Ground state

No electronic excitations, no electron-phonon coupling, etc

3. Classical approx. for nuclei movement

No proton transfer, atom tunneling, etc

On the good side – big systems, long evolution times, nice parallelization and scaling

TD Schrödinger eq. – 3 atoms

MD – 10^9 atoms, up to milliseconds

Mechanics: main concepts

Newton laws

$$\vec{F} = m * \vec{a}$$

$$\vec{a} = \frac{d\vec{v}}{dt} = \frac{d^2\vec{r}}{dt^2}$$

$$\vec{F} = -grad(U)$$

$$\vec{p} = m\vec{v}$$

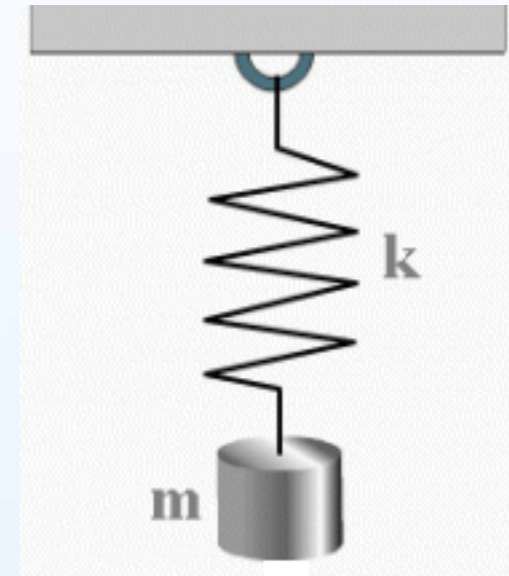
U – potential energy

T – kinetic energy

H – full energy,
Hamiltonian

$$H = T + U$$

$$\frac{d^2\vec{r}}{dt^2} = \vec{F}/m$$



$$U = \frac{kx^2}{2}$$

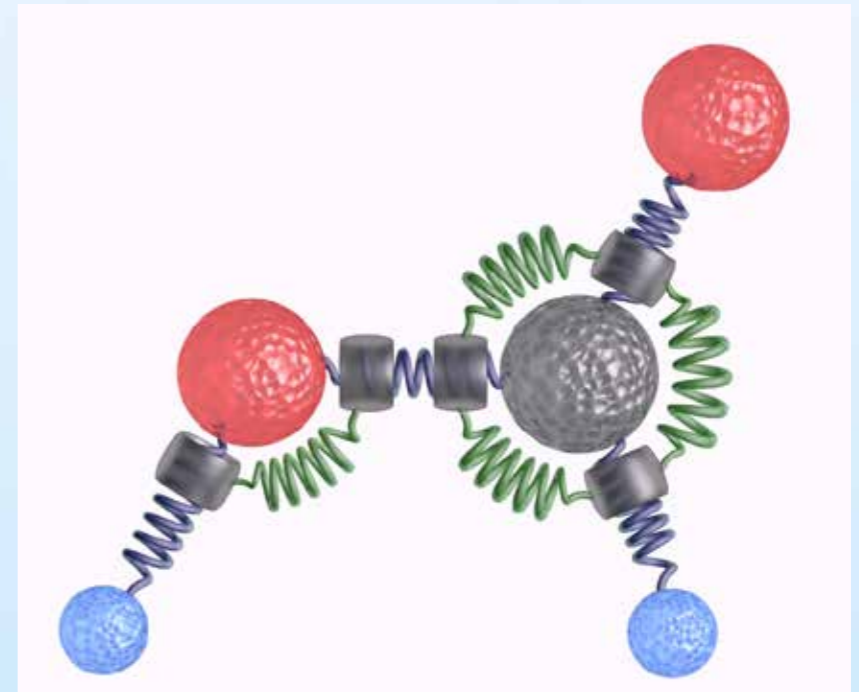
$$T = \frac{mv^2}{2}$$

PES, force field

$$\frac{d^2\vec{r}}{dt^2} = -grad(U) \quad U = E_e(\vec{r}) \quad H_e \chi = E_e \chi$$

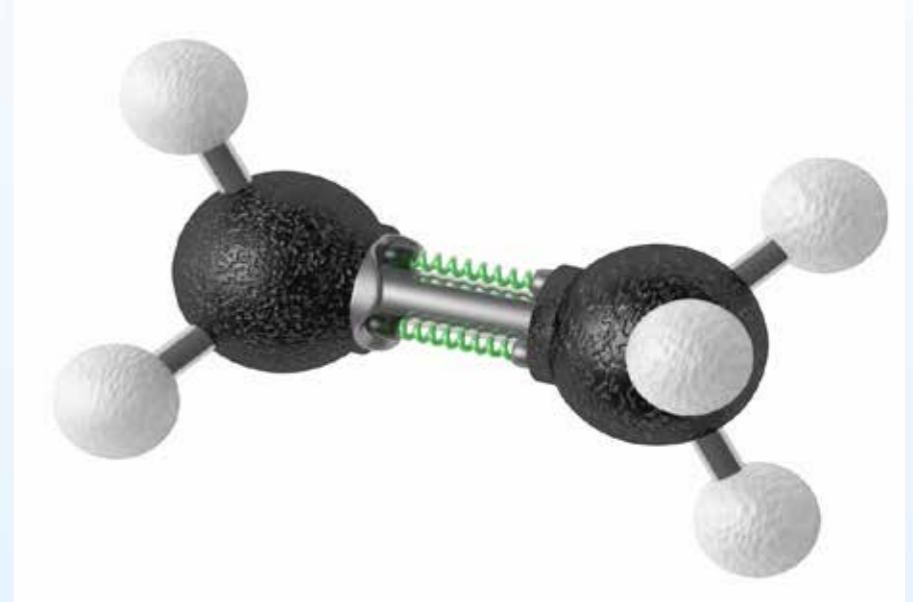
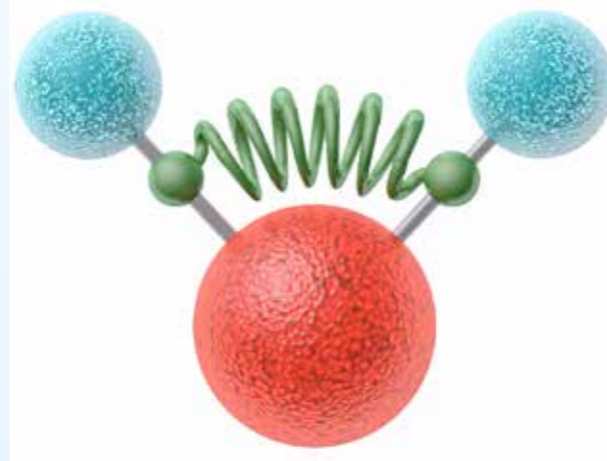
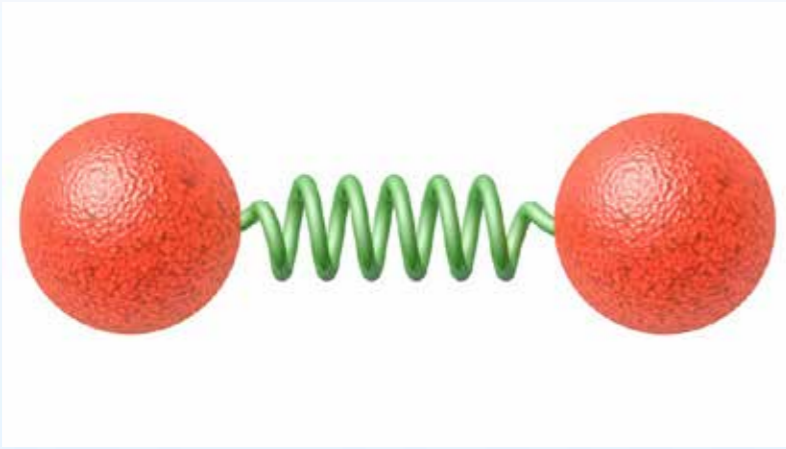
Empirical potential energy functions
– force fields

1. Knowledge of chemical structure
2. Combination of typical simple interaction terms
3. Often pair wise additive
4. Empirical parameters fit to experimental or ab initio data



Typical force field

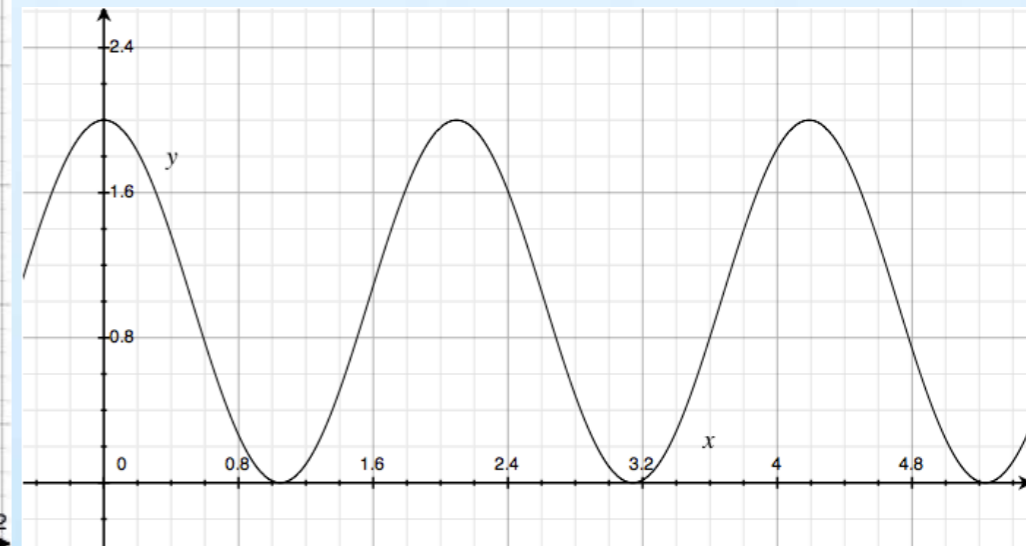
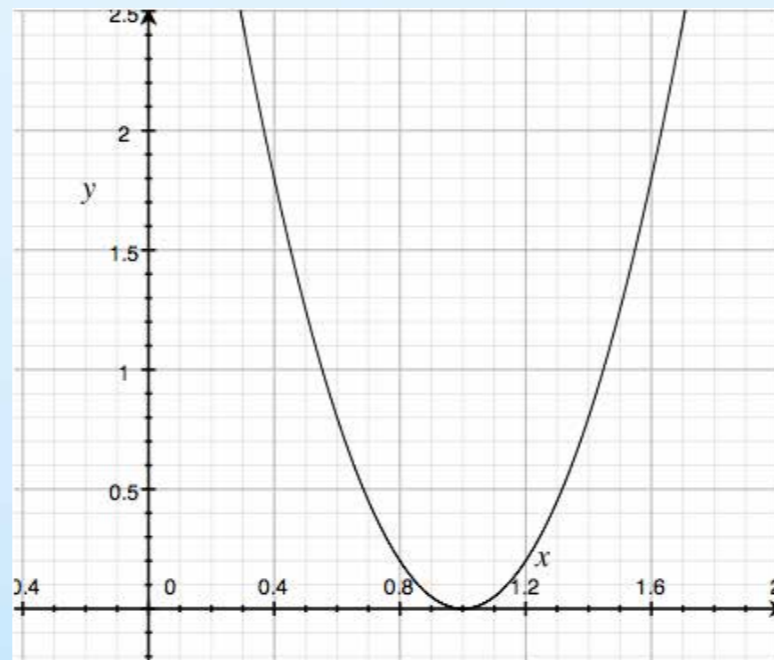
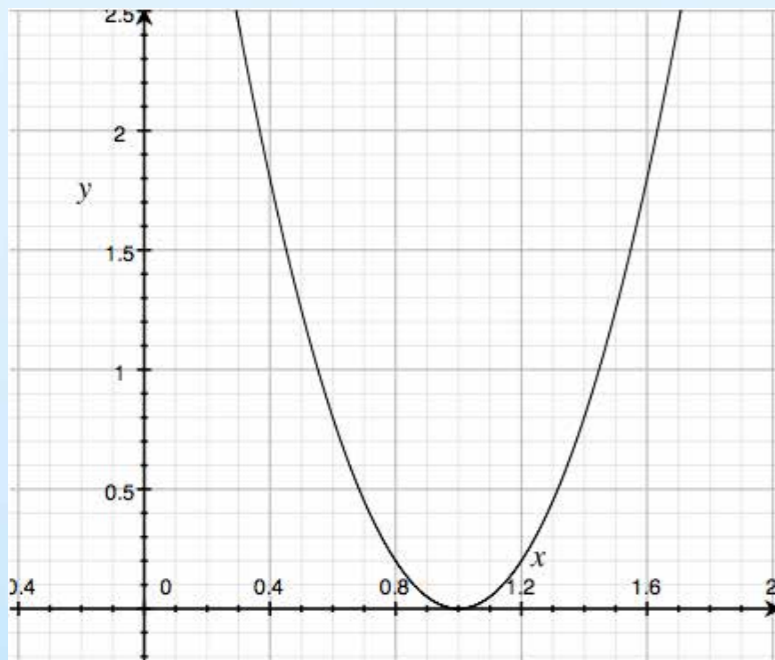
Bonded terms



$$U_{bond} = \frac{k(l - l_0)^2}{2}$$

$$U_{ang} = \frac{k(\theta - \theta_0)^2}{2}$$

$$U_{tor} = V[1 + \cos(n\phi - \phi_0)]$$

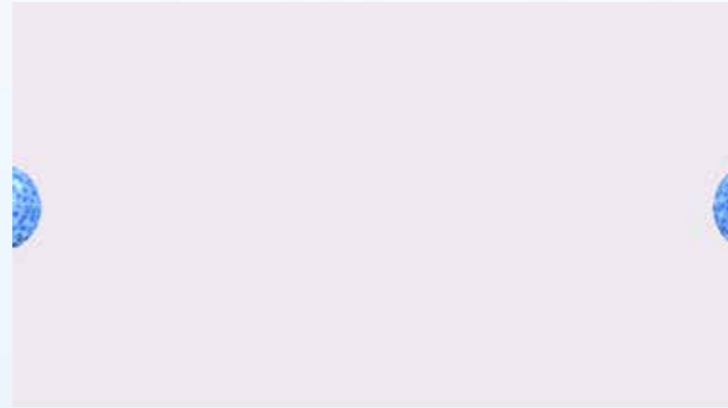
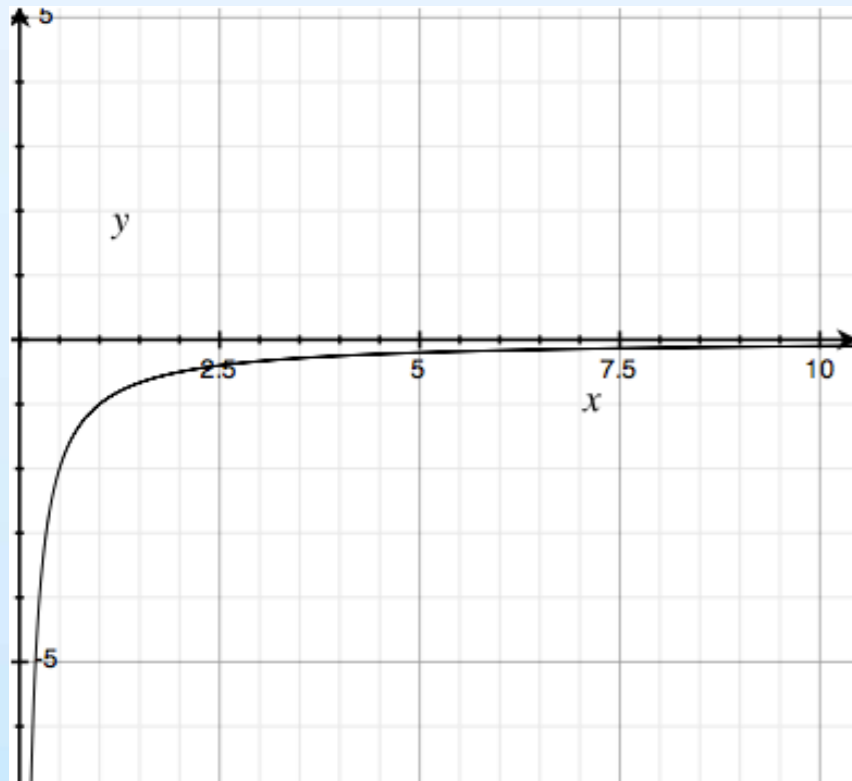


Typical force field

Non-bonded terms

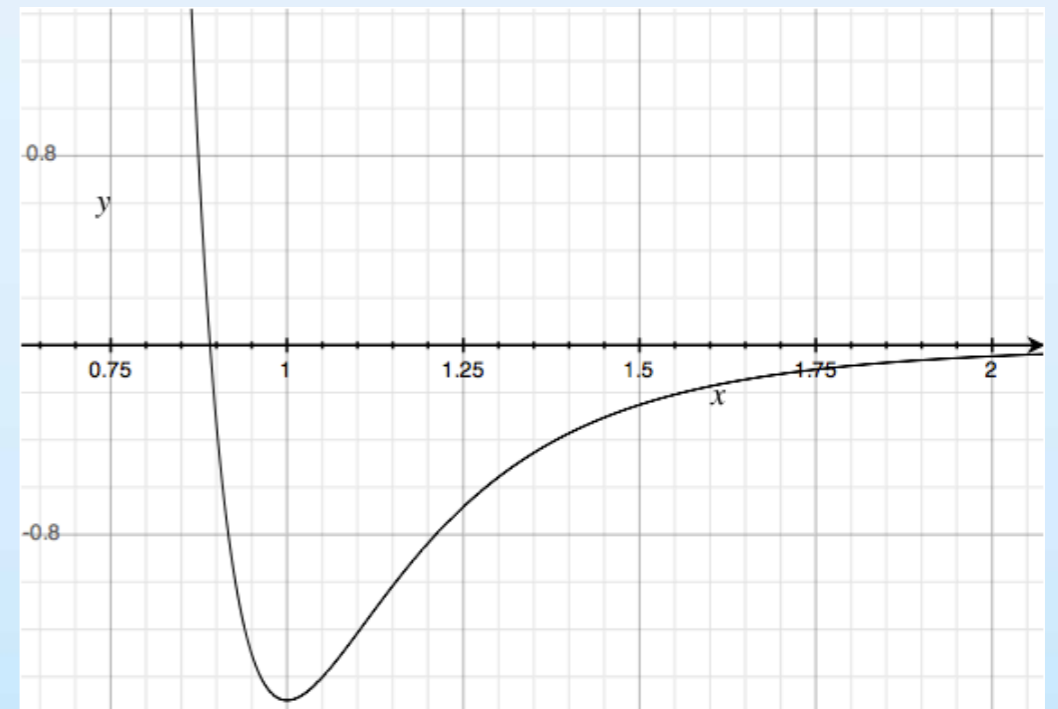
Coulomb

$$U_q = \frac{q_i q_j}{r}$$

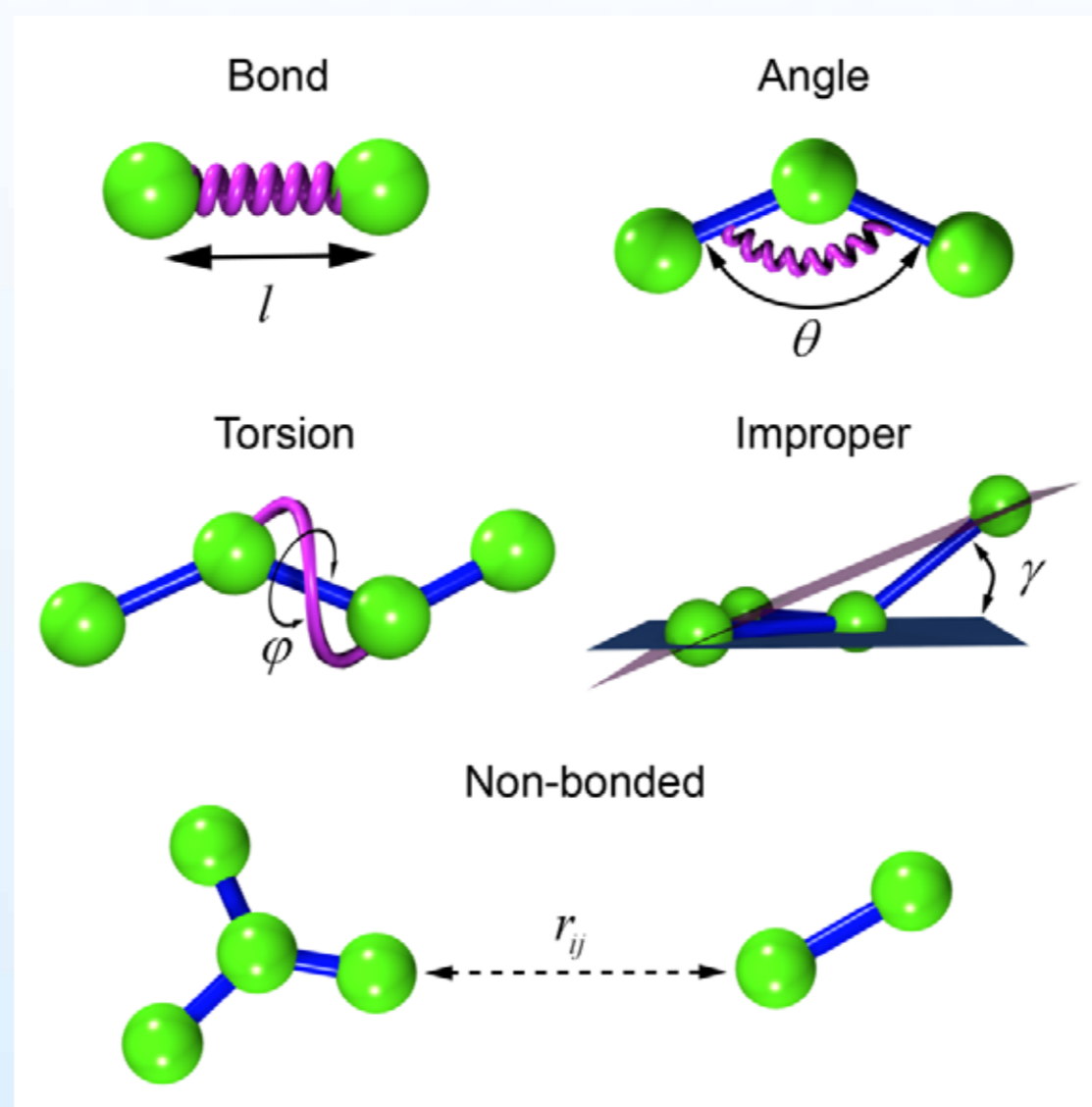


Lennard-Jones

$$U_{LJ} = 4\epsilon \left[\left(\frac{\sigma}{r} \right)^{12} - \left(\frac{\sigma}{r} \right)^6 \right]$$



Typical force field



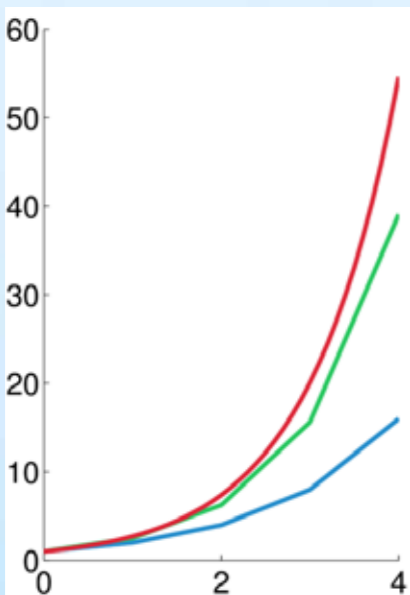
$$\begin{aligned}
 U(\{\vec{r}_i\}) = & \sum_{\text{bonds}} \frac{1}{2} k_b (l - l_0)^2 + \sum_{\text{angles}} \frac{1}{2} k_\theta (\theta - \theta_0)^2 + \sum_{\text{torsions}} \frac{1}{2} V_n [1 + \cos(n\varphi - \varphi_0)] \\
 & + \sum_{\text{impropes}} \frac{1}{2} k_\gamma (\gamma - \gamma_0)^2 + \sum_{j=1}^{N-1} \sum_{i=j+1}^N \left\{ 4\epsilon_{ij} \left[\left(\frac{\sigma_{ij}}{r_{ij}} \right)^{12} - \left(\frac{\sigma_{ij}}{r_{ij}} \right)^6 \right] + \frac{q_i q_j}{4\pi\epsilon_0 r_{ij}} \right\} f_{ij}
 \end{aligned}$$

Integrating equations of motion

$$\begin{cases} \frac{d\vec{r}_i}{dt} = \vec{v}_i \\ \frac{d\vec{v}_i}{dt} = \vec{F}_i/m_i \end{cases} \implies \begin{cases} \frac{\vec{r}_i(t+\Delta t) - \vec{r}_i(t)}{\Delta t} = \vec{v}_i \\ \frac{\vec{v}_i(t+\Delta t) - \vec{v}_i(t)}{\Delta t} = \vec{F}_i(t)/m_i \end{cases}$$

Euler Scheme

$$\begin{cases} \vec{r}_i(t) \\ \vec{v}_i(t) \end{cases} \implies \begin{cases} \vec{r}_i(t + \Delta t) = \vec{r}_i(t) + \vec{v}_i(t) * \Delta t \\ \vec{v}_i(t + \Delta t) = \vec{v}_i(t) + \frac{\vec{F}_i(t)}{m_i} * \Delta t \end{cases}$$



- First order algorithm
- Instable
- High energy drifts
- No time reversibility
- Not area preserving (non-symplectic)

Integrating equations of motion

Störmer-Verlet scheme

$$\frac{\frac{\vec{r}_i(t+\Delta t) - \vec{r}_i(t)}{\Delta t} - \frac{\vec{r}_i(t) - \vec{r}_i(t-\Delta t)}{\Delta t}}{\Delta t} = \frac{\vec{F}_i(t)}{m_i}$$

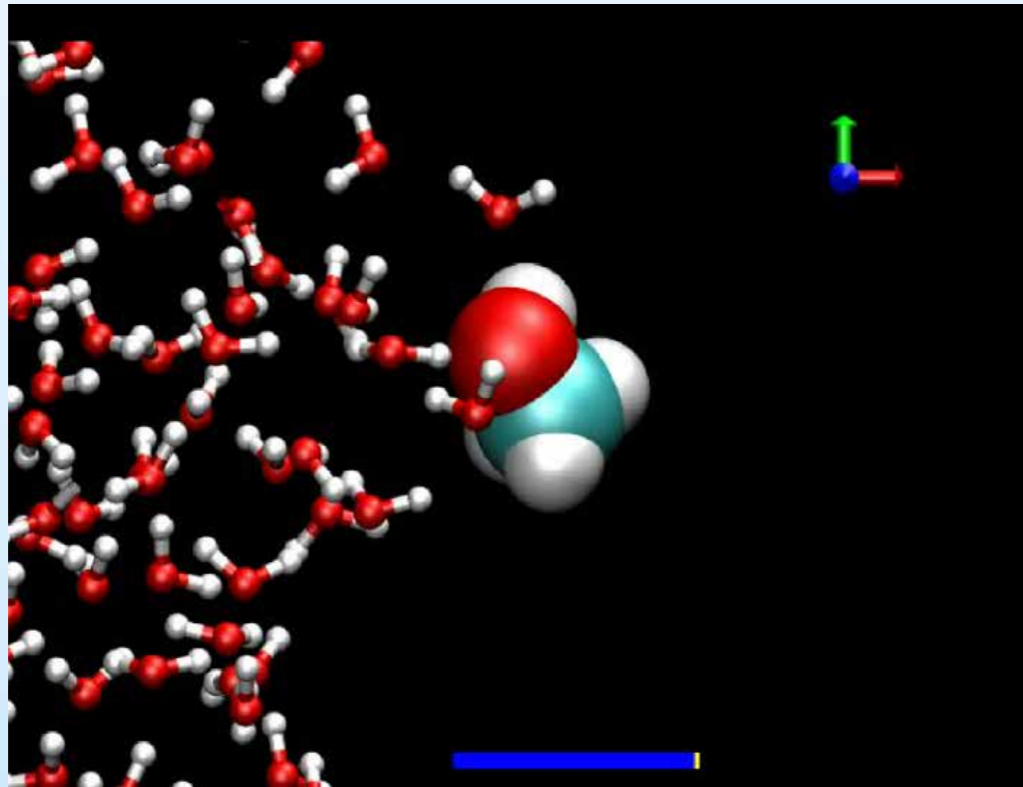
$$\vec{r}_i(t + \Delta t) = 2\vec{r}_i(t) - \vec{r}_i(t - \Delta t) + \frac{\vec{F}_i(t)}{m_i} * \Delta t^2$$

- No long term energy drift (except for rounding errors)
- Nice stability
- Time reversible
- Global error of order $O(\Delta t^2)$

Integrating equations of motion

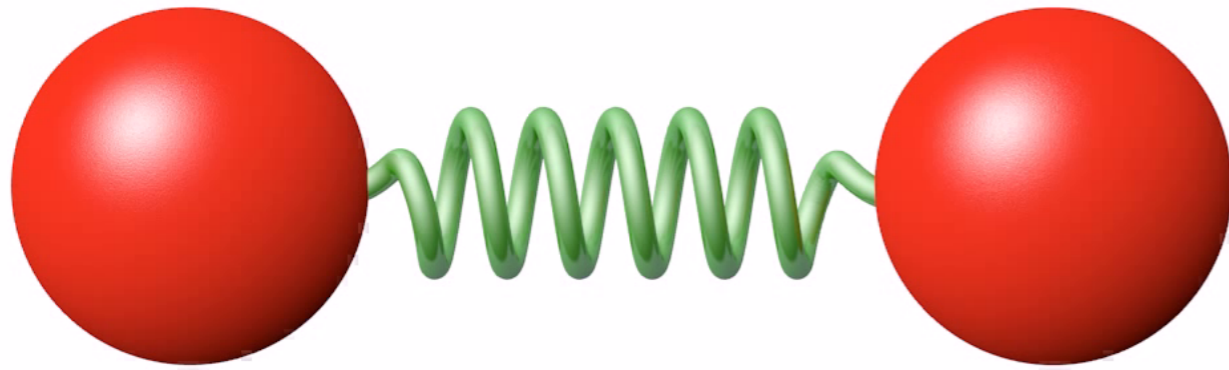
Velocity verlet scheme (leap frog)

$$\begin{cases} \vec{r}_i(t + \Delta t) = \vec{r}_i(t) + \vec{v}_i(t) \Delta t + \frac{1}{2} \vec{a}_i \Delta t^2 \\ \vec{v}_i(t + \Delta t) = \vec{v}_i(t) + \frac{\vec{a}_i(t) + \vec{a}_i(t + \Delta t)}{2} \Delta t \end{cases}$$

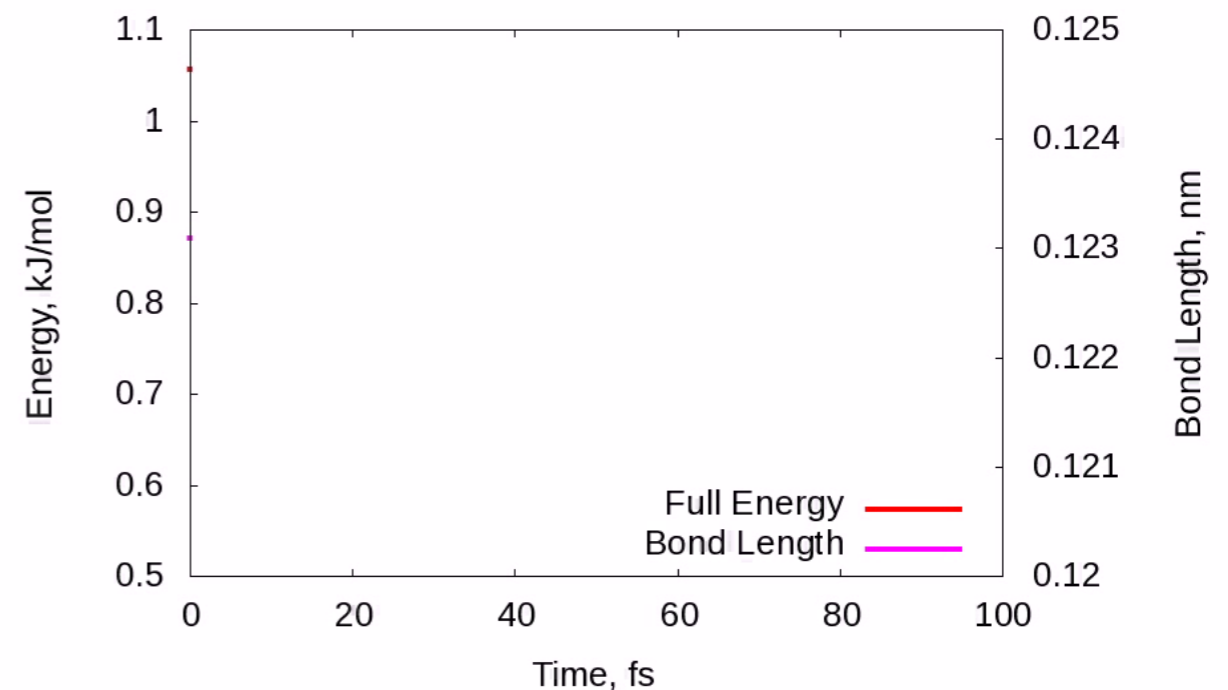
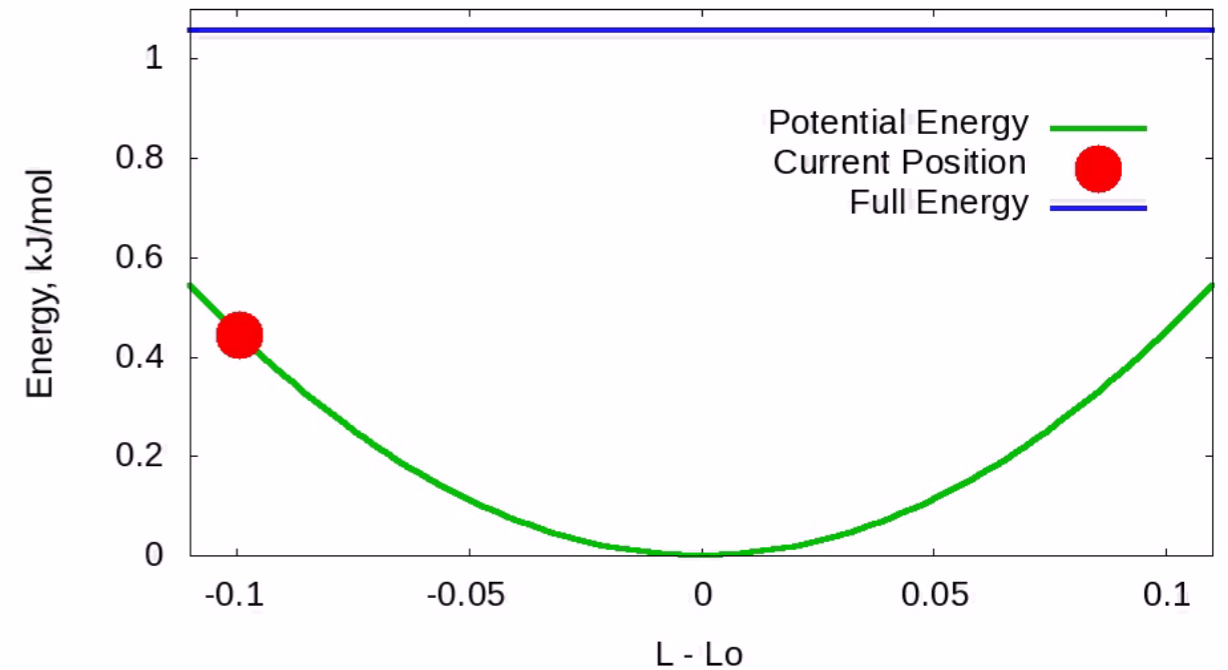


Example of MD
simulation

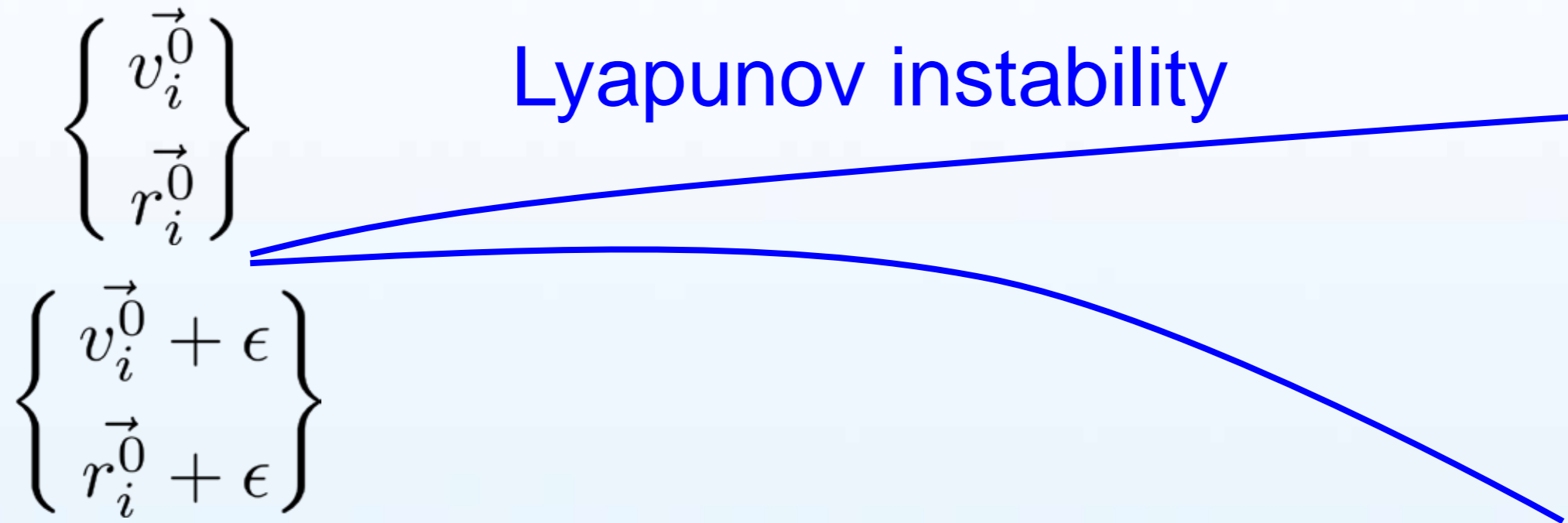
Integrating equations of motion



Oxygen molecule
NVE simulation
Leap-frog integrator
Integration step 1 fs

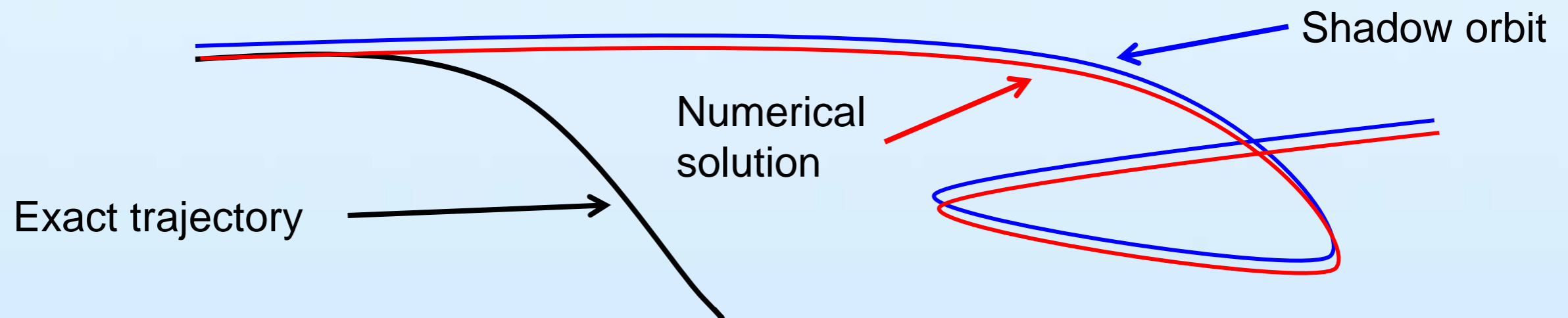


MD picture vs reality?



How can we know that we generate correct results in MD?

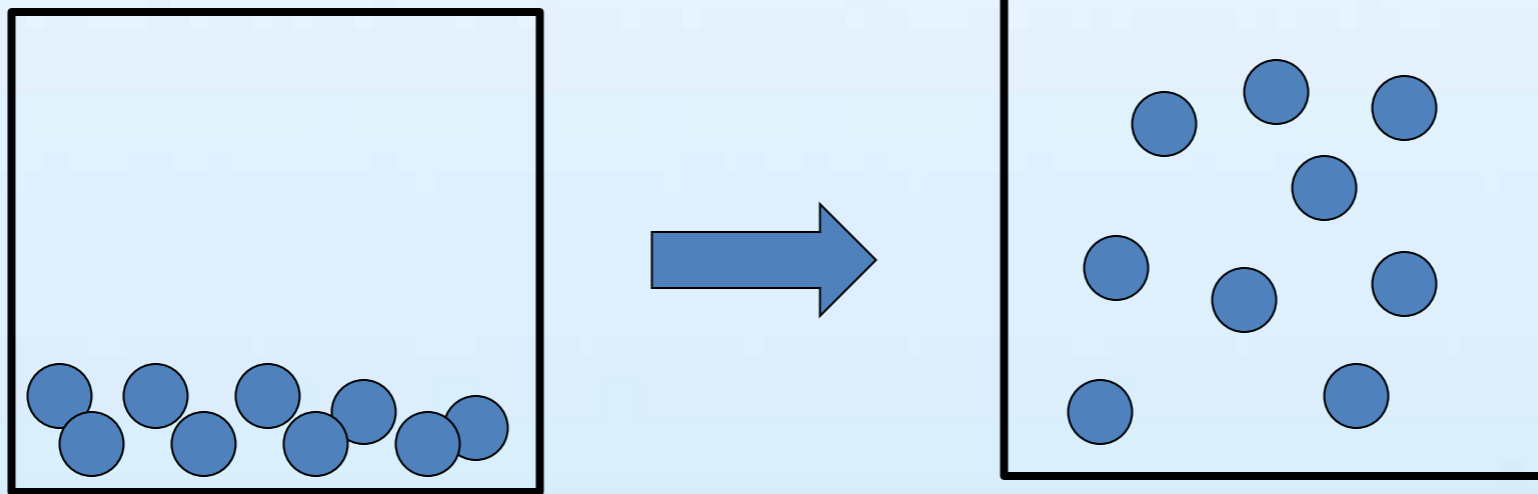
- We can not know for sure, but there is hope ...



A *shadow orbit* is an exact trajectory for the system, that starts from a slightly different initial point

MD picture vs reality?

The aim of MD is not to predict precisely what will happen to a system that has been prepared in precisely known initial conditions – but to predict average behavior of the system prepared in an initial state about which we know something but not everything.



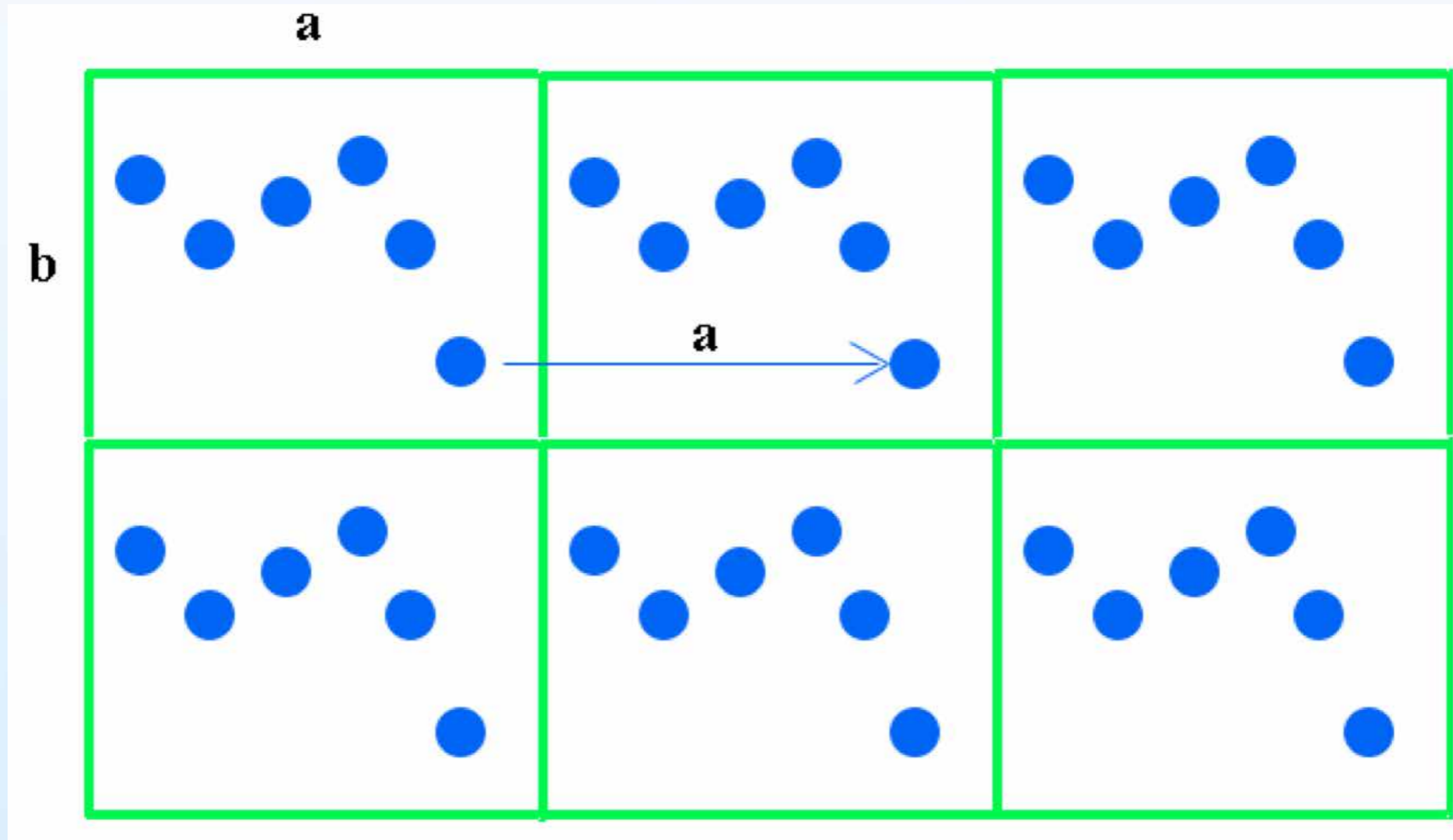
MD tricks: level 1

Needed almost in every simulation

- Periodic boundary conditions
- Interaction cut off
- Constant temperature/pressure simulations
- Parallelization and HPC computing

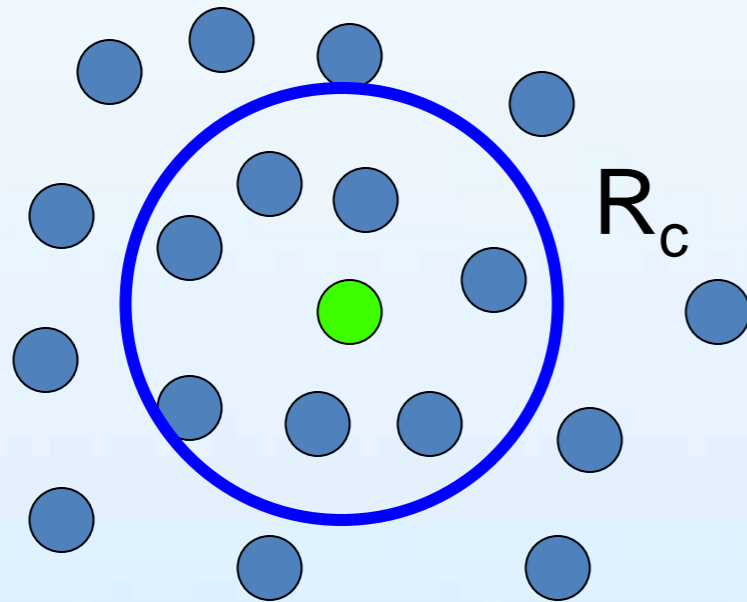
Periodic boundary conditions

A trade off against boundary effects

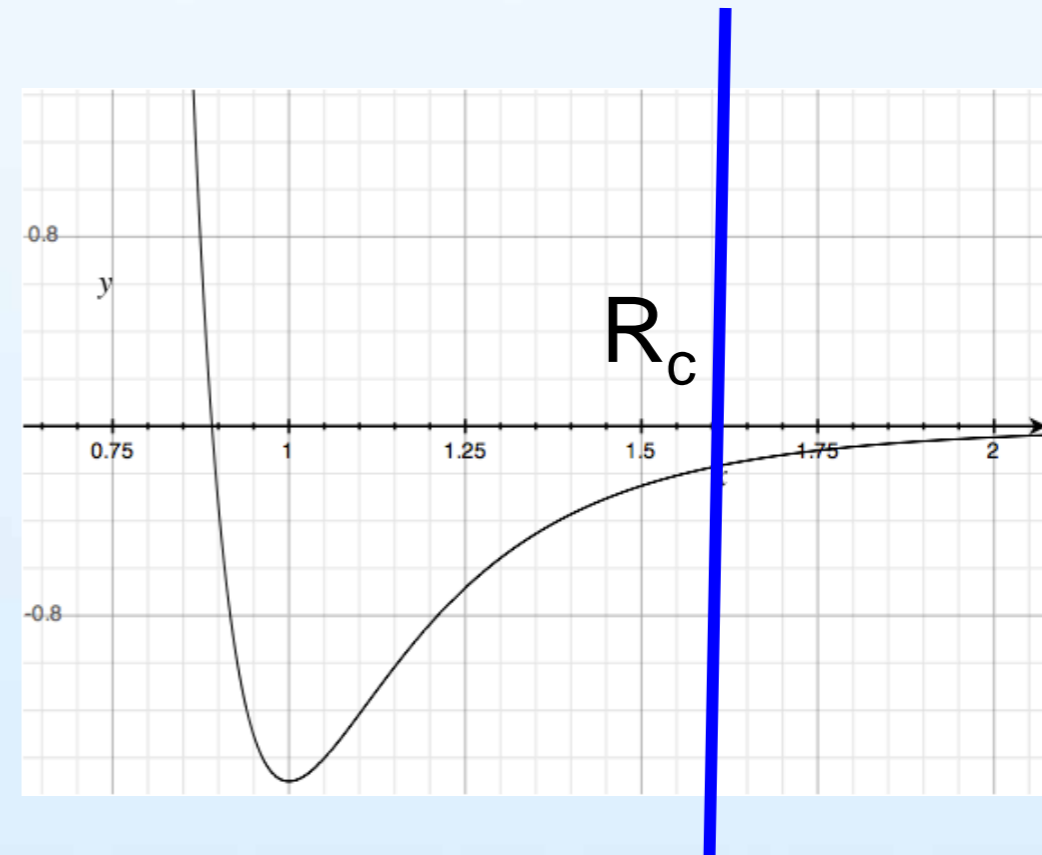


$$\vec{r}_{klm} = \vec{r} + k * \vec{a} + l * \vec{b} + m * \vec{c}$$

Cutoff schemes



Simple cutoff radius

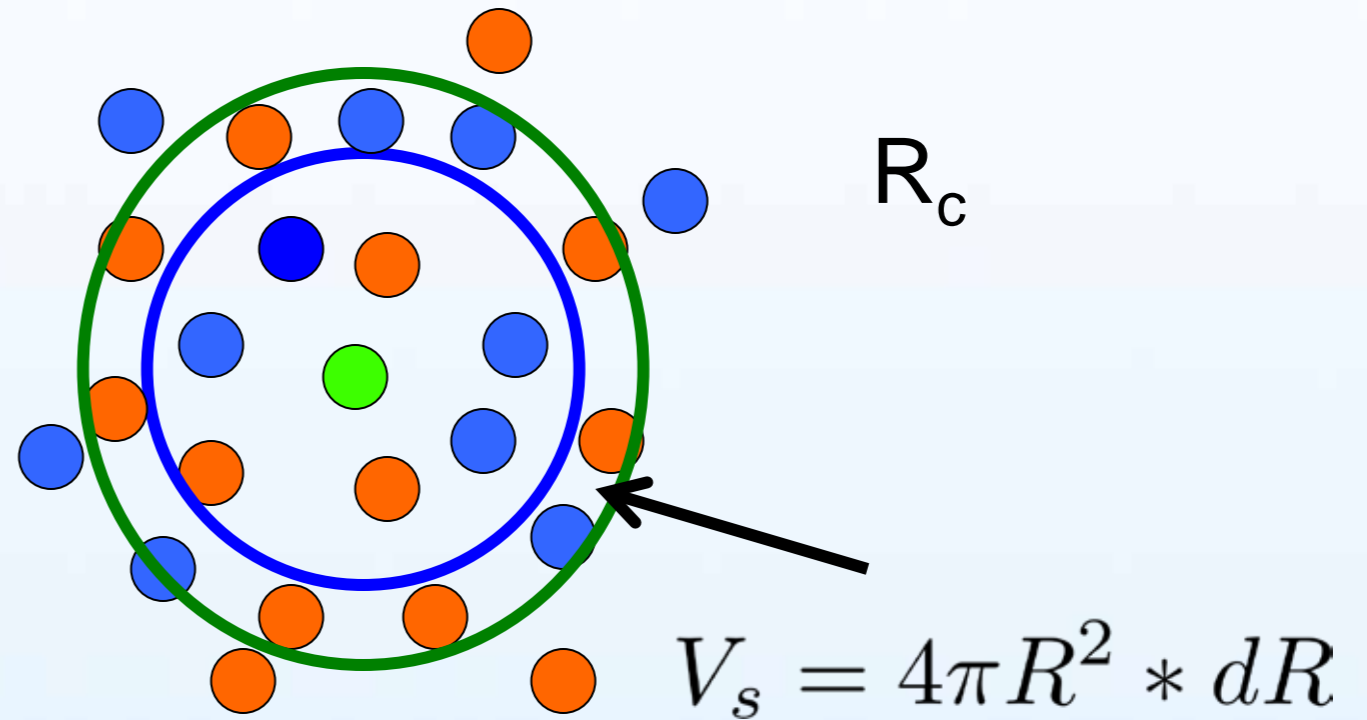


Electrostatics

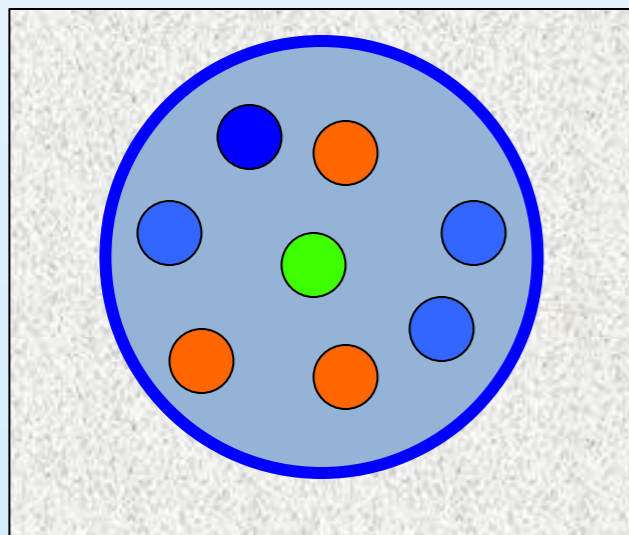
$$U_q = \frac{q_i q_j}{r}$$

Coulomb potential

$$\phi(r_i) = q \sum_{j+}^{\infty} \frac{1}{|r_i - r_{j+}|} - q \sum_{j-}^{\infty} \frac{1}{|r_i - r_{j-}|}$$

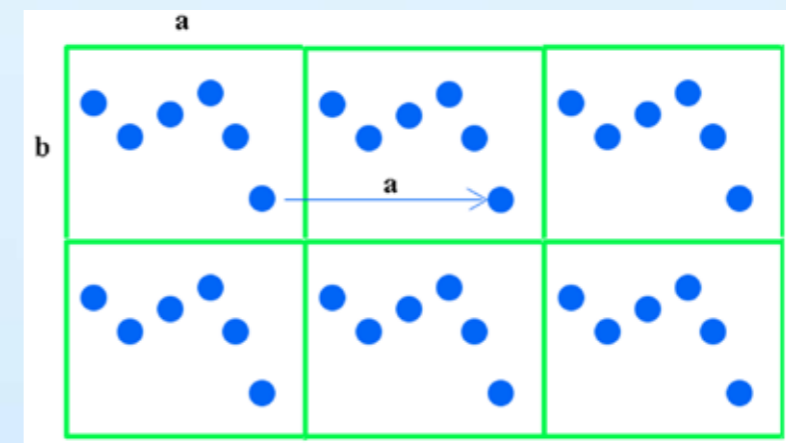


Reaction field



Make use of Poisson-Boltzmann equation

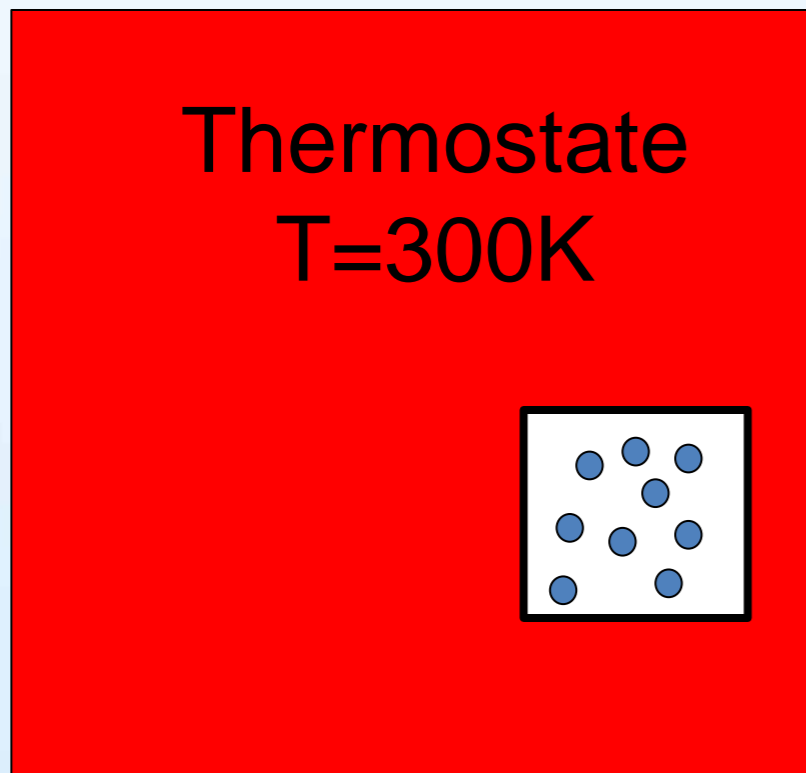
Ewald summation



Make use of periodicity and reciprocal space summation

Constant temperature simulations

- Why use temperature coupling?
- -Compensate energy drift
- -NVT ensemble



$$P_{microstate} \propto e^{-\frac{H}{kT}}$$

Gibbs distribution

Two possibilities:

- 1) Add stochastic interactions
- 2) Use modified equations of motion

Constant temperature simulations 2

Stochastic approach

Andersen thermostat
-random collision with virtual particles

$$P(t) = \nu e^{-\nu t}$$

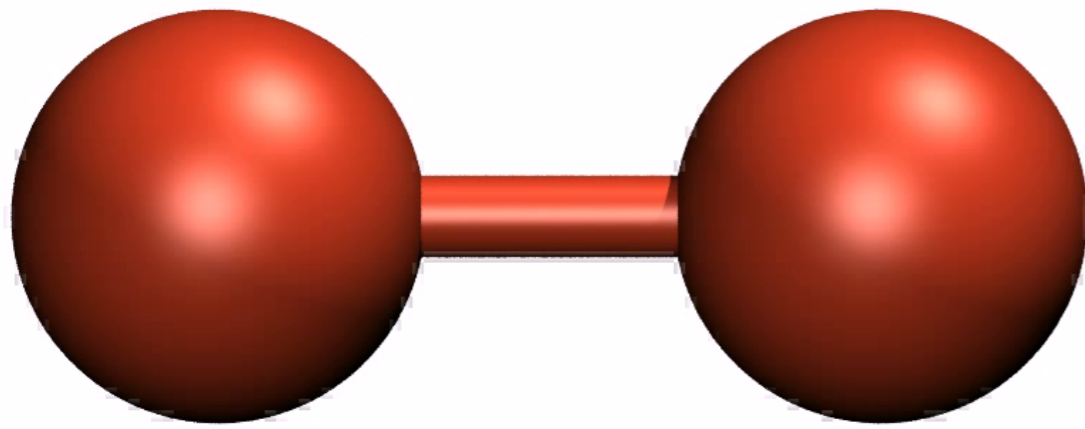
Langevin stochastic dynamics
-add friction and stochastic term

$$m \frac{d^2 \vec{r}_i}{dt^2} = \vec{F}_i - \lambda \frac{d\vec{r}_i}{dt} + \boldsymbol{\eta}(t)$$

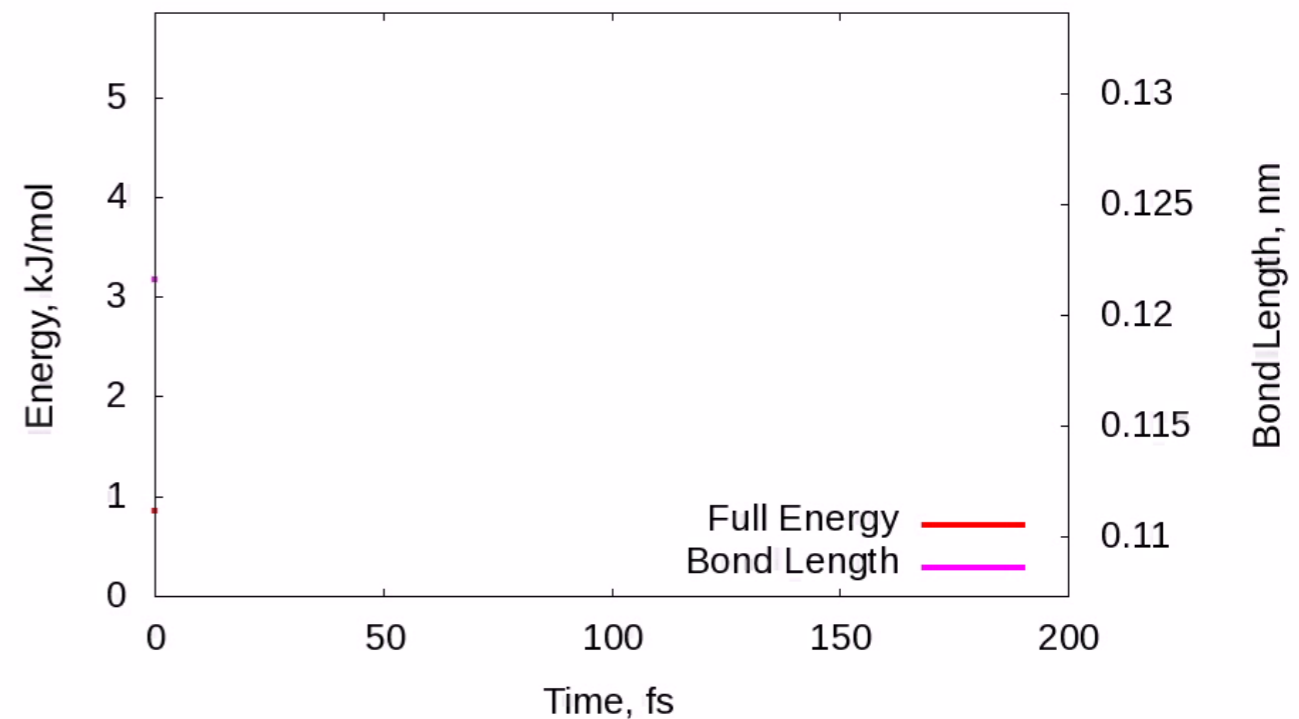
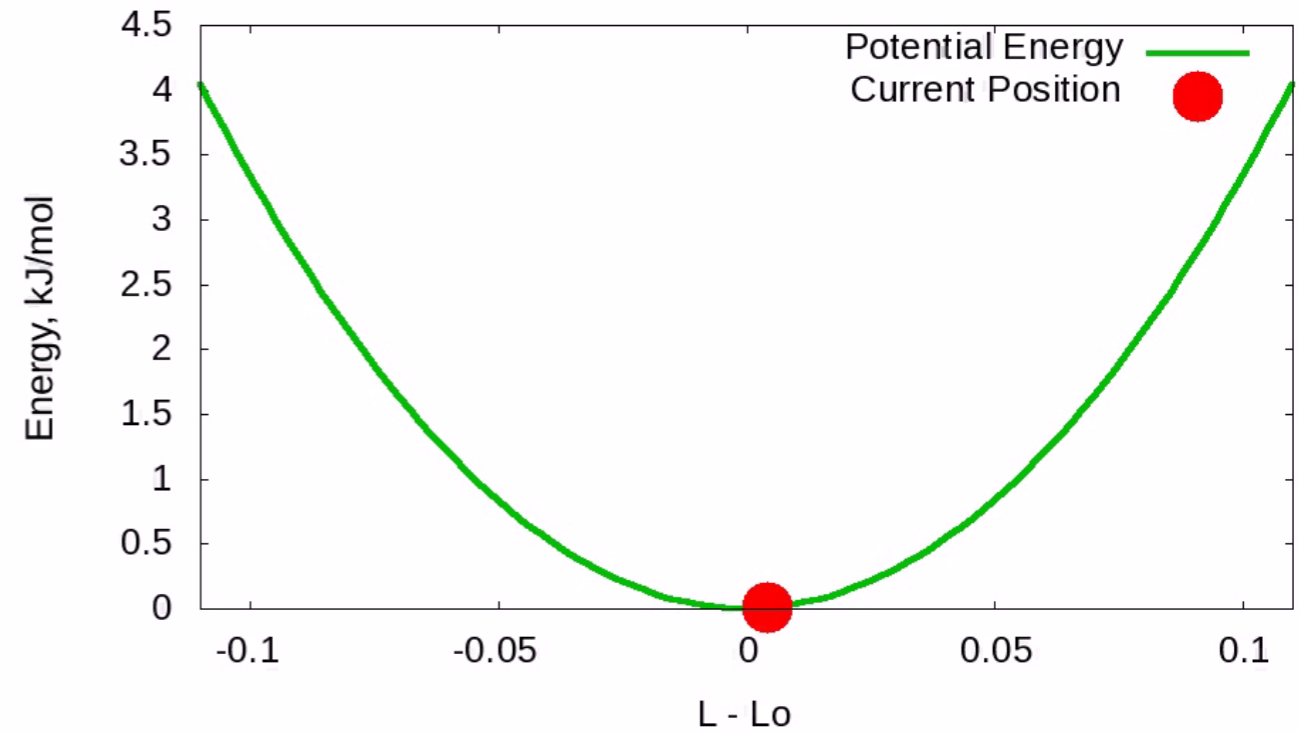
$$\langle \eta_i(t) \eta_j(t') \rangle = 2\lambda k_B T \delta_{i,j} \delta(t - t')$$

Constant temperature simulations 2

Langevin dynamics



Oxygen molecule
Stochastic dynamics simulation



Constant temperature simulations 3

Berendsen thermostat

$$\frac{kT}{2}(3N - N_c) = \sum_{i=1}^N m_i \vec{v}_i^2 \quad \text{-instantaneous temperature}$$

Modify equations of motion

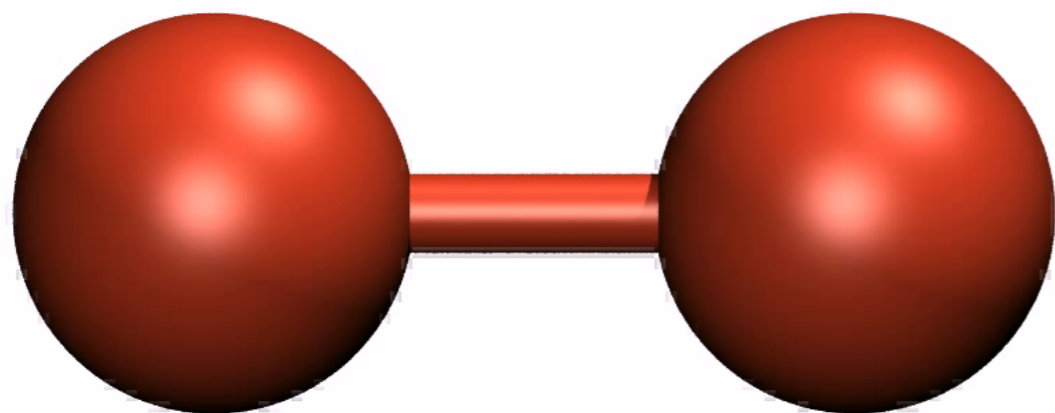
-re-scale the velocities of particles, multiply by

$$\lambda = \sqrt{T_0/T(t)}$$

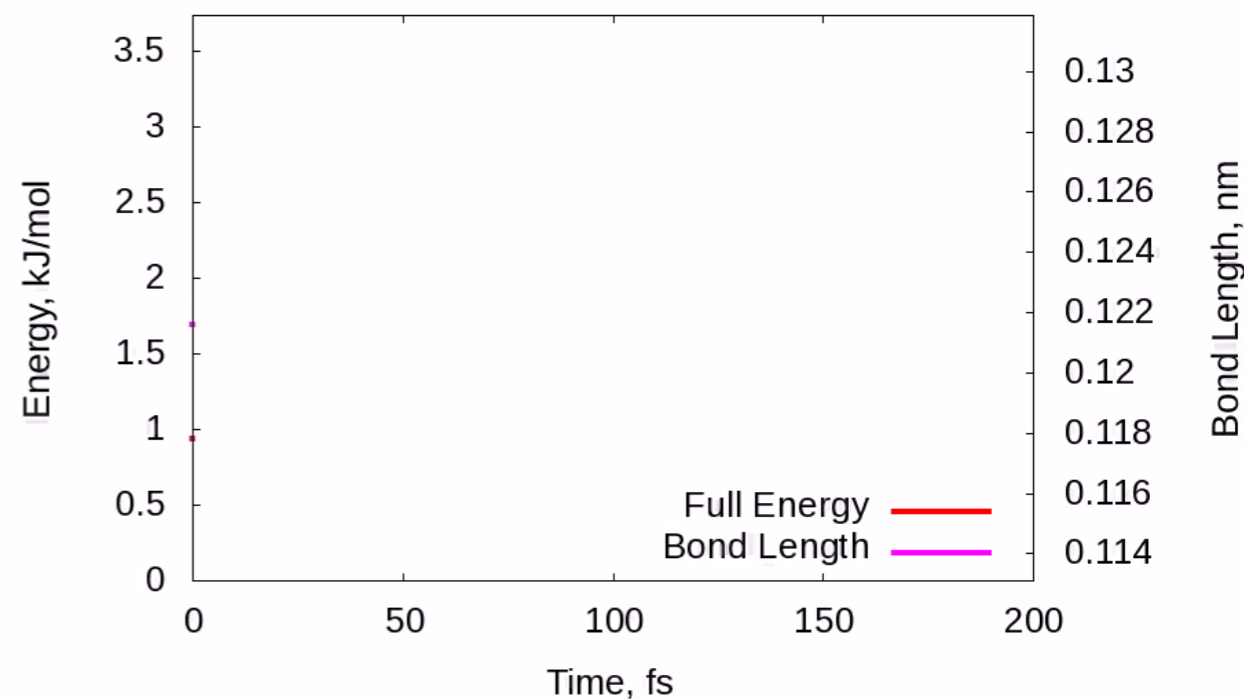
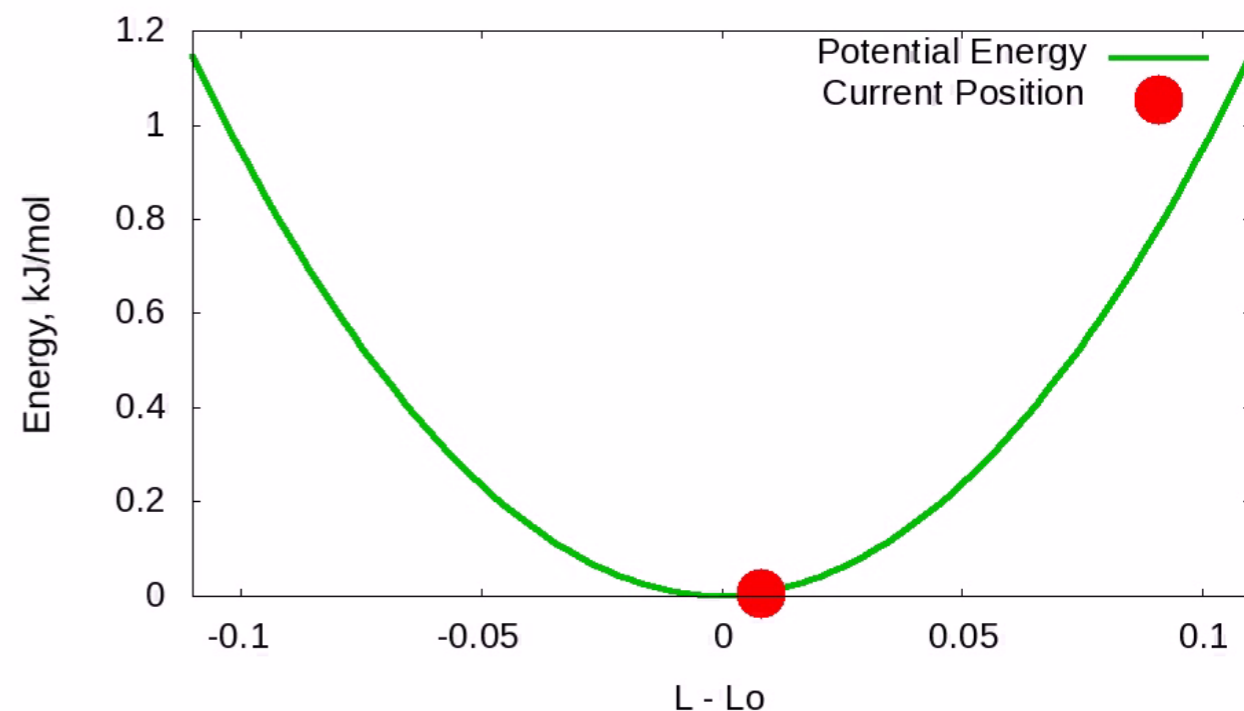
(!) incorrect ensemble, flying ice cube effect

Constant temperature simulations 3

Berendsen thermostat



Oxygen molecule
NVT simulation
Leap-Frog integrator
Integration step 1 fs
Berendsen thermostat



Constant temperature simulations 3

Nose-Hoover thermostat

Extended system approach

-add a “bath particle”

-modify equations of motion, Non-Hamiltonian dynamics

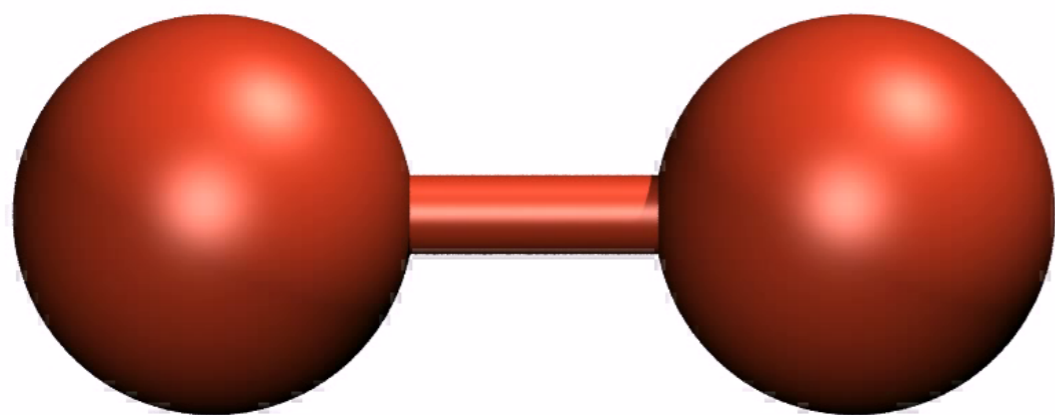
-gives NVT ensemble if:

1) No external forces, 2) there is only one conservation law, 3) the system is ergodic

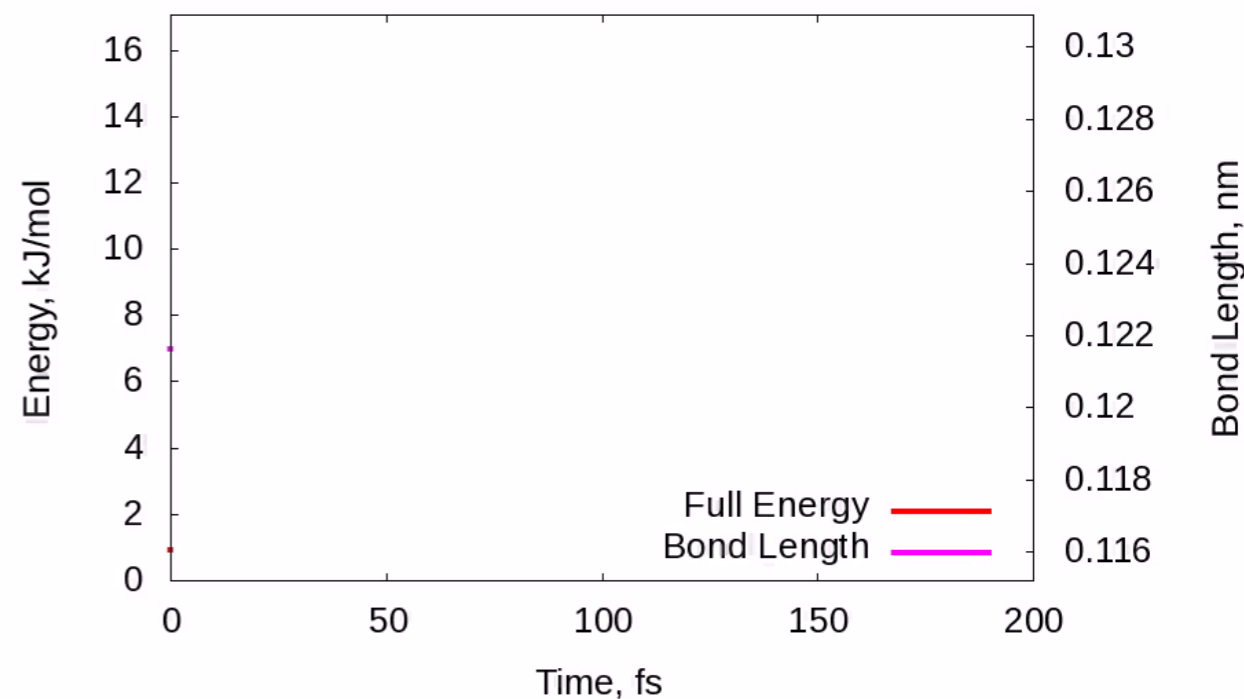
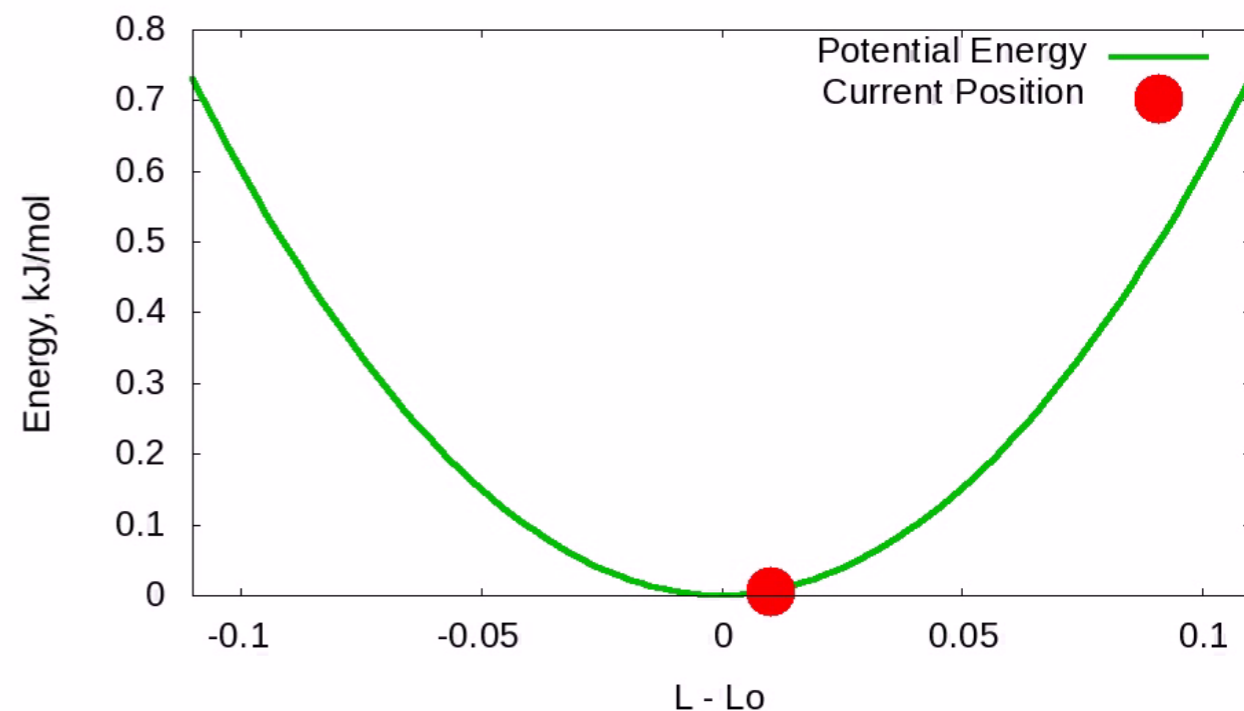
$$\left\{ \begin{array}{l} \frac{d^2 \vec{r}_i}{dt^2} = \frac{F_i}{m_i} - \frac{d\xi}{dt} \frac{d\vec{r}_i}{dt} \\ \frac{d^2 \xi}{dt^2} = \frac{N_f k}{Q} [T - T_0] \end{array} \right. \quad E_{NH} = E_k + U + \frac{1}{2} Q \dot{\xi}^2 + N_f k T \xi$$

Constant temperature simulations 3

Nose-Hoover thermostat



Oxygen molecule
NVT simulation
Verlet integrator
Integration step 1 fs
Nose-Hoover thermostat



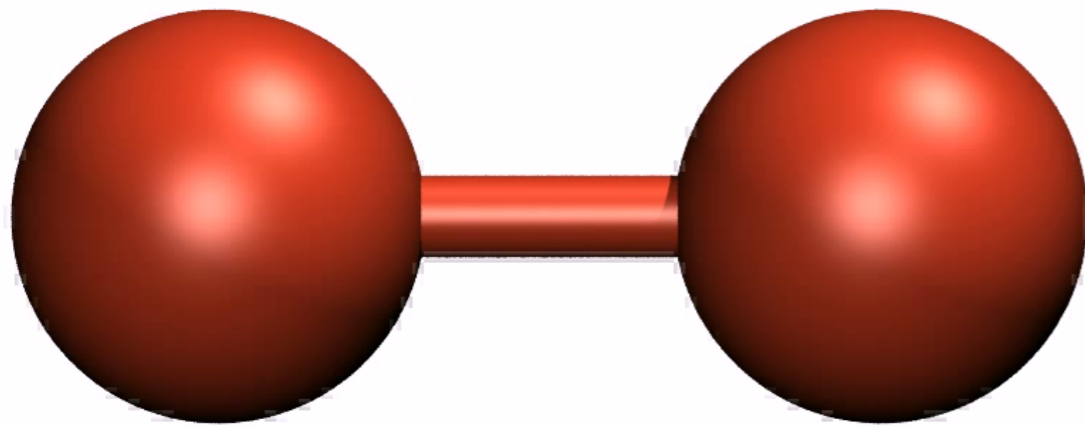
Constant temperature simulations 3

Nose-Hoover chains

$$\begin{aligned}\dot{\mathbf{r}}_i &= \frac{\mathbf{p}_i}{m_i} \\ \dot{\mathbf{p}}_i &= \mathbf{F}_i - \mathbf{p}_i \dot{\xi}_1 \\ \ddot{\xi}_1 &= \frac{1}{Q_1} \left[\sum_{i=1}^N m_i \mathbf{v}_i^2 - N_f k_B T_{set} \right] - \dot{\xi}_1 \dot{\xi}_2 \\ \ddot{\xi}_j &= \frac{1}{Q_j} \left[Q_{j-1} \dot{\xi}_{j-1}^2 - k_B T_{set} \right] - \dot{\xi}_j \dot{\xi}_{j+1} \quad , j = 2, \dots, M-1 \\ \ddot{\xi}_M &= \frac{1}{Q_M} \left[Q_{M-1} \dot{\xi}_{M-1}^2 - k_B T_{set} \right] \quad ,\end{aligned}$$

Constant temperature simulations 3

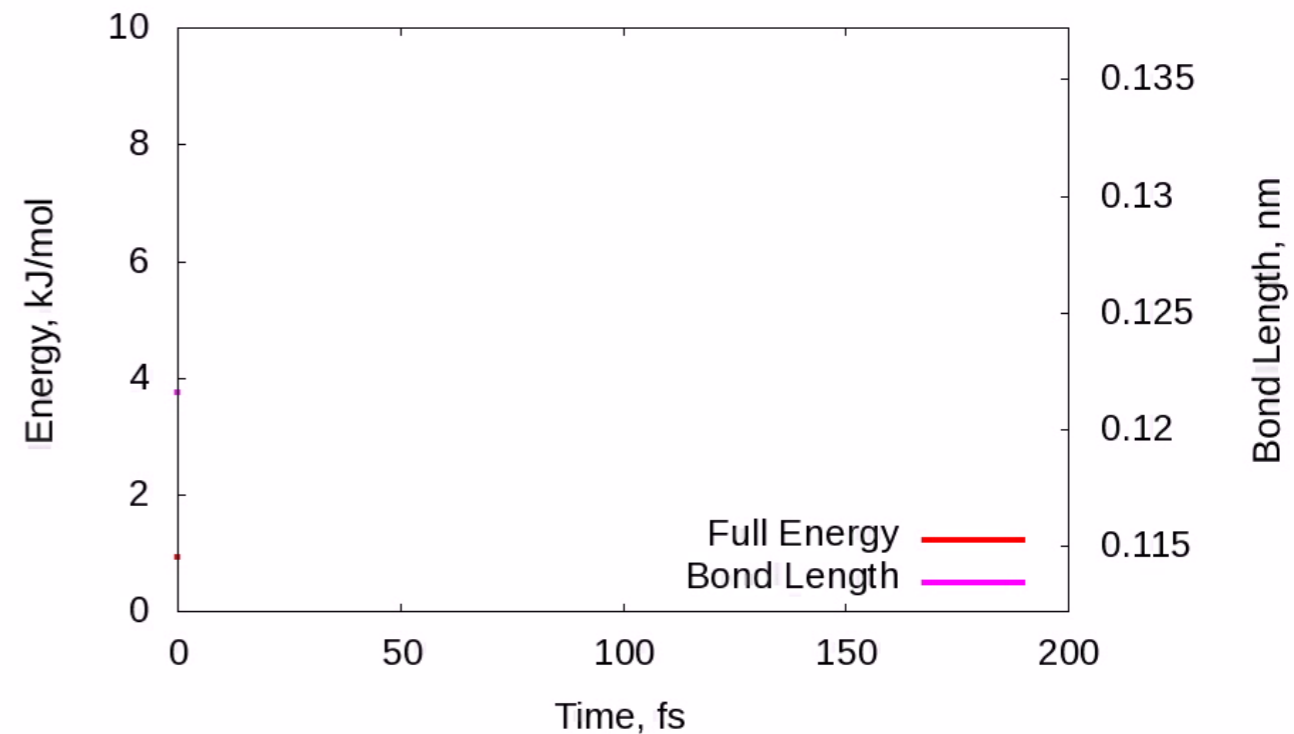
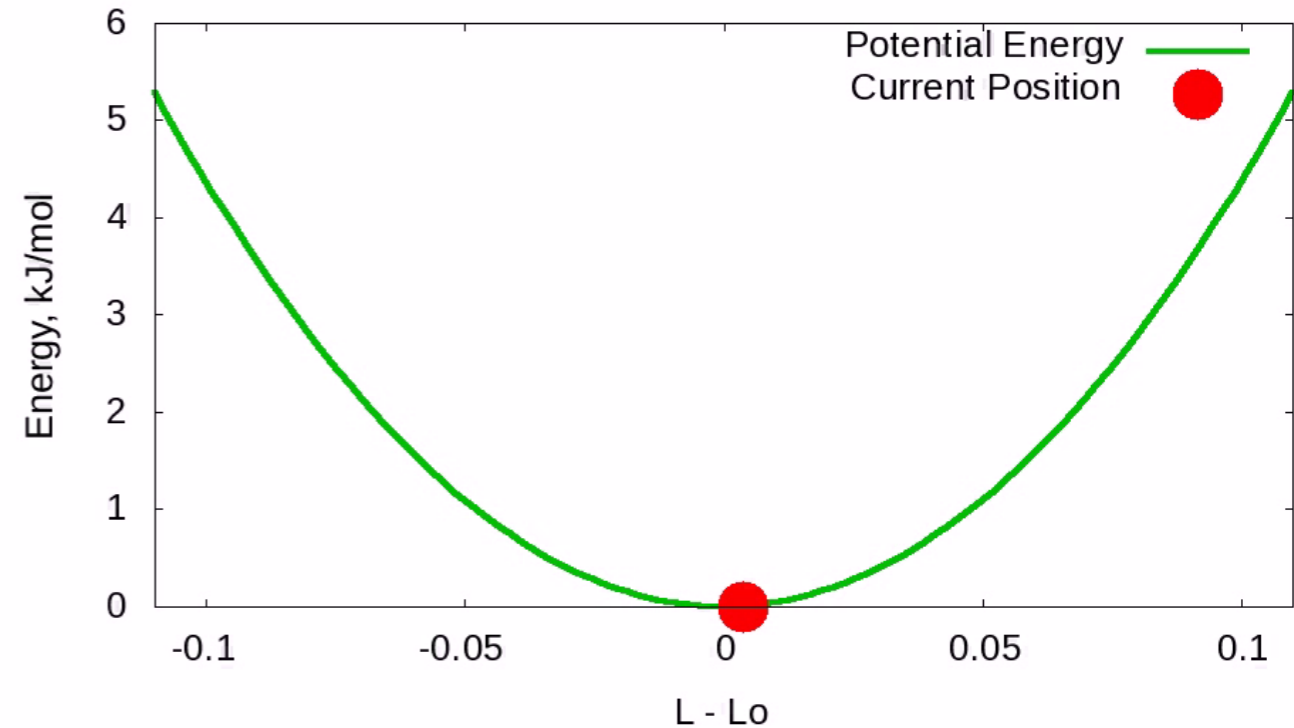
Nose-Hoover chains



Oxygen molecule
NVT simulation
Verlet integrator

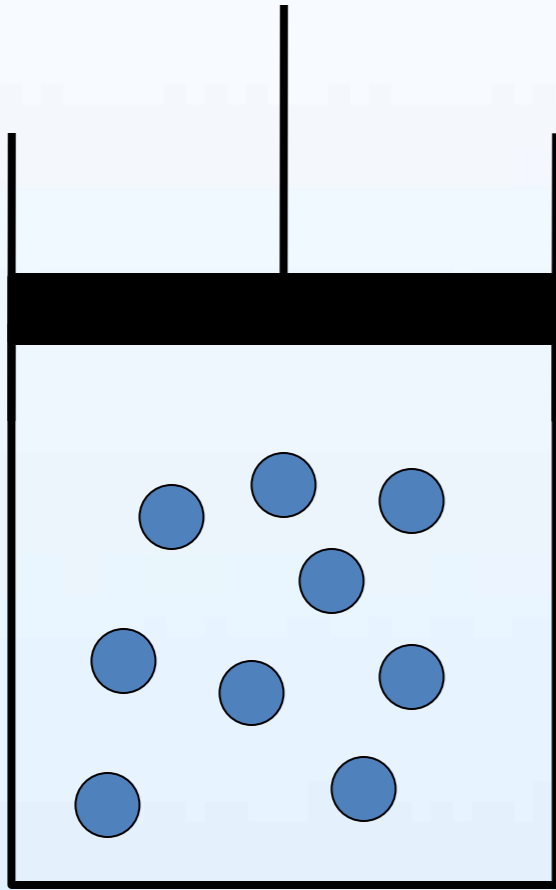
Integration step 1 fs

Nose-Hoover + chains thermostat



Constant pressure simulations

Berendsen Barostat



$$P = nkT + vir/V$$

$$vir = \frac{1}{3} \sum \vec{f}_{ij} \vec{r}_{ij}$$

$$\begin{cases} \vec{r}_i \implies \mu \vec{r}_i \\ L \implies \mu L \end{cases}$$

NPT-ensemble

$$\mu = \left[1 - \frac{\Delta t}{\tau_p} (P - P_0) \right]$$

Computational costs

Initial coordinates

Computational time

Force calculation

$$\mathbf{F}_i = -\frac{\partial V}{\partial \mathbf{r}_i}$$

$$\mathbf{F}_i = \sum_j \mathbf{F}_{ij}$$

Coordinate update

$$\frac{d^2 \mathbf{r}_i}{dt^2} = \frac{\mathbf{F}_i}{m_i}$$

Data output

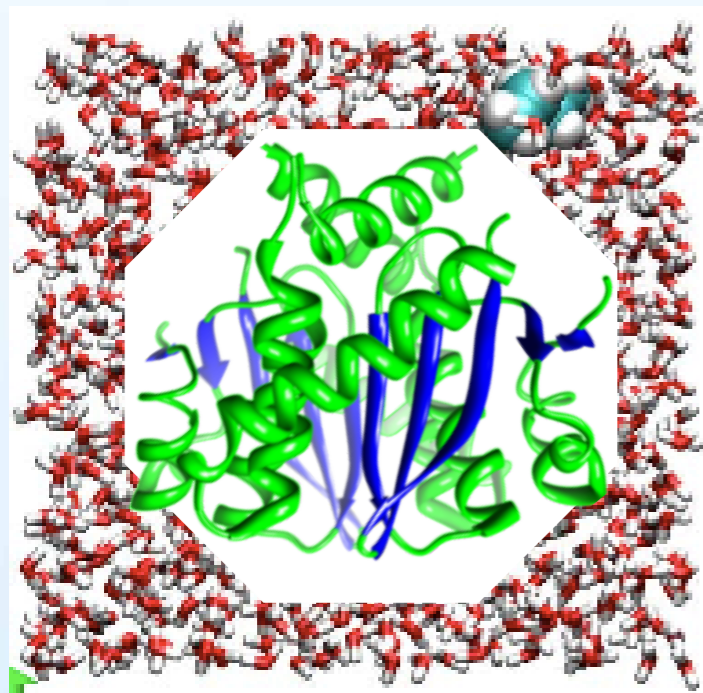
$$\begin{aligned} & \sum_{\theta} K_{\theta} (\theta - \theta_0)^2 + \\ & \sum_{\phi} K_{\phi} (1 + \cos(n\phi - \phi_0)) \\ & \sum_{ij} \epsilon \left[\left(\frac{\sigma}{r} \right)^{12} - \left(\frac{\sigma}{r} \right)^6 \right] + \\ & \sum_{ij} \frac{q_i q_j}{r_{ij}} \end{aligned}$$

} ~2 %

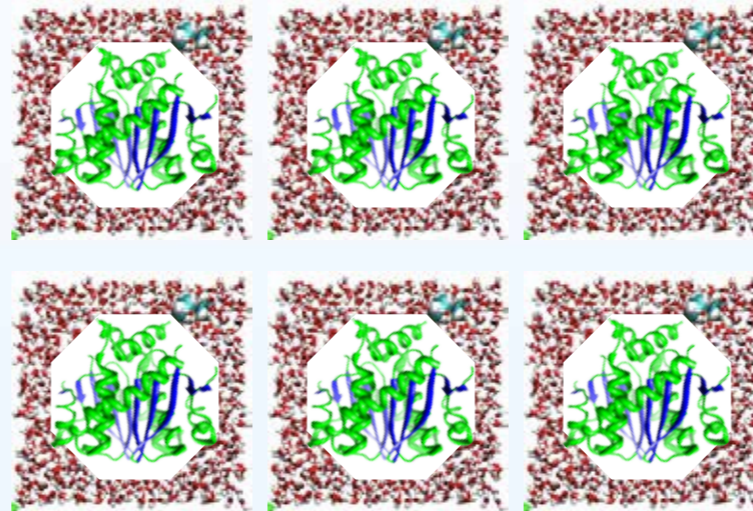
} ~90 %

} ~0.1 %

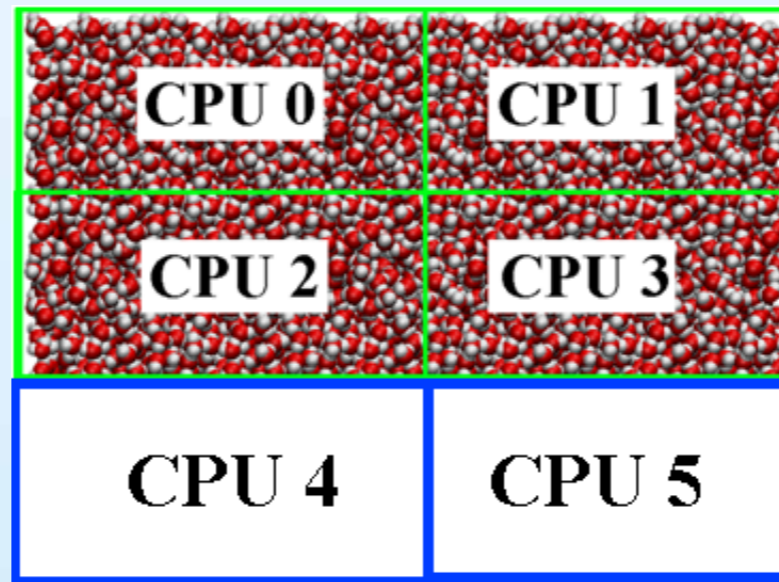
Parallel computing



System under study



Multiple replica calculation



Domain/task decomposition

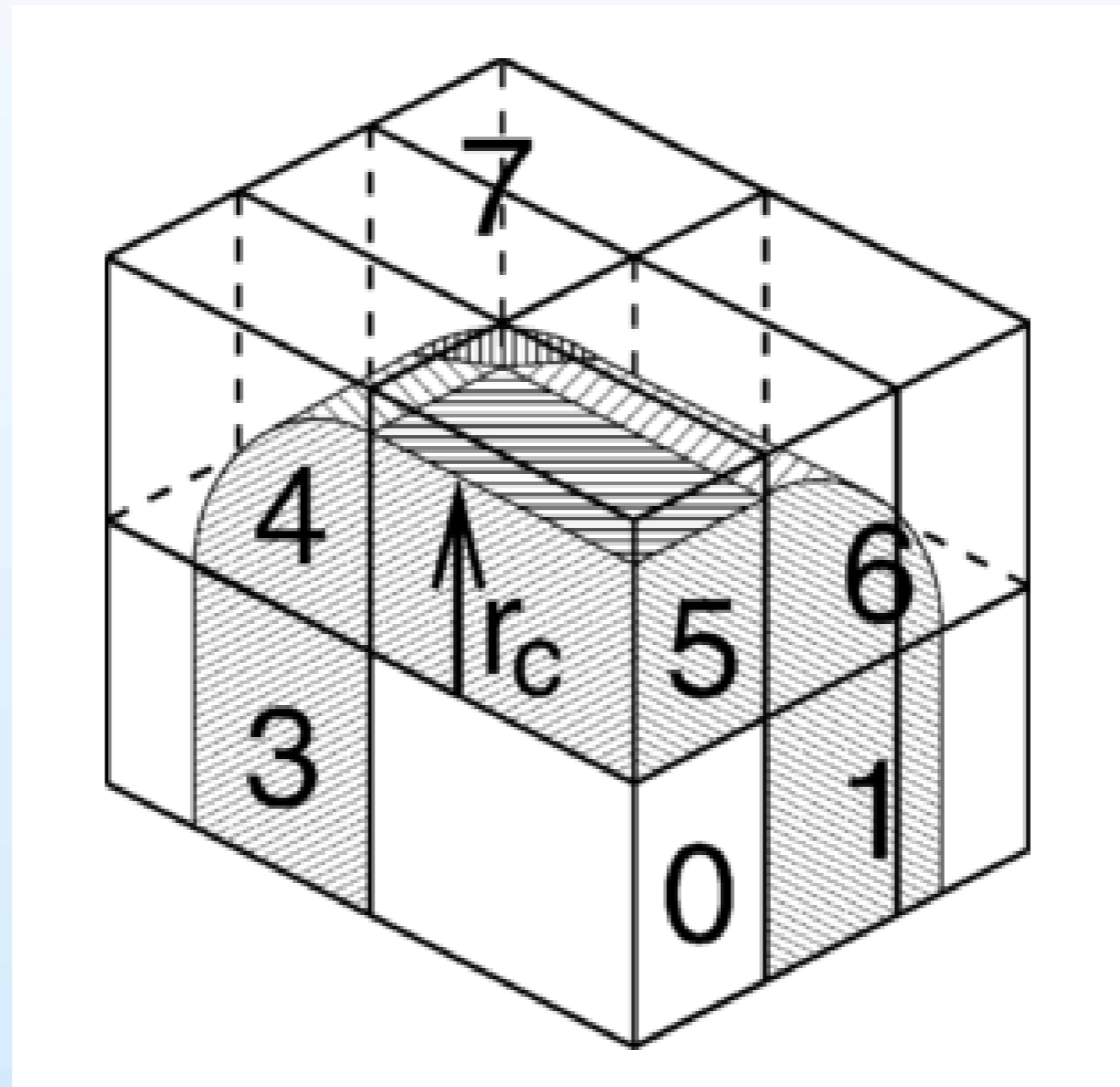


GRIDs, distributed computing, etc

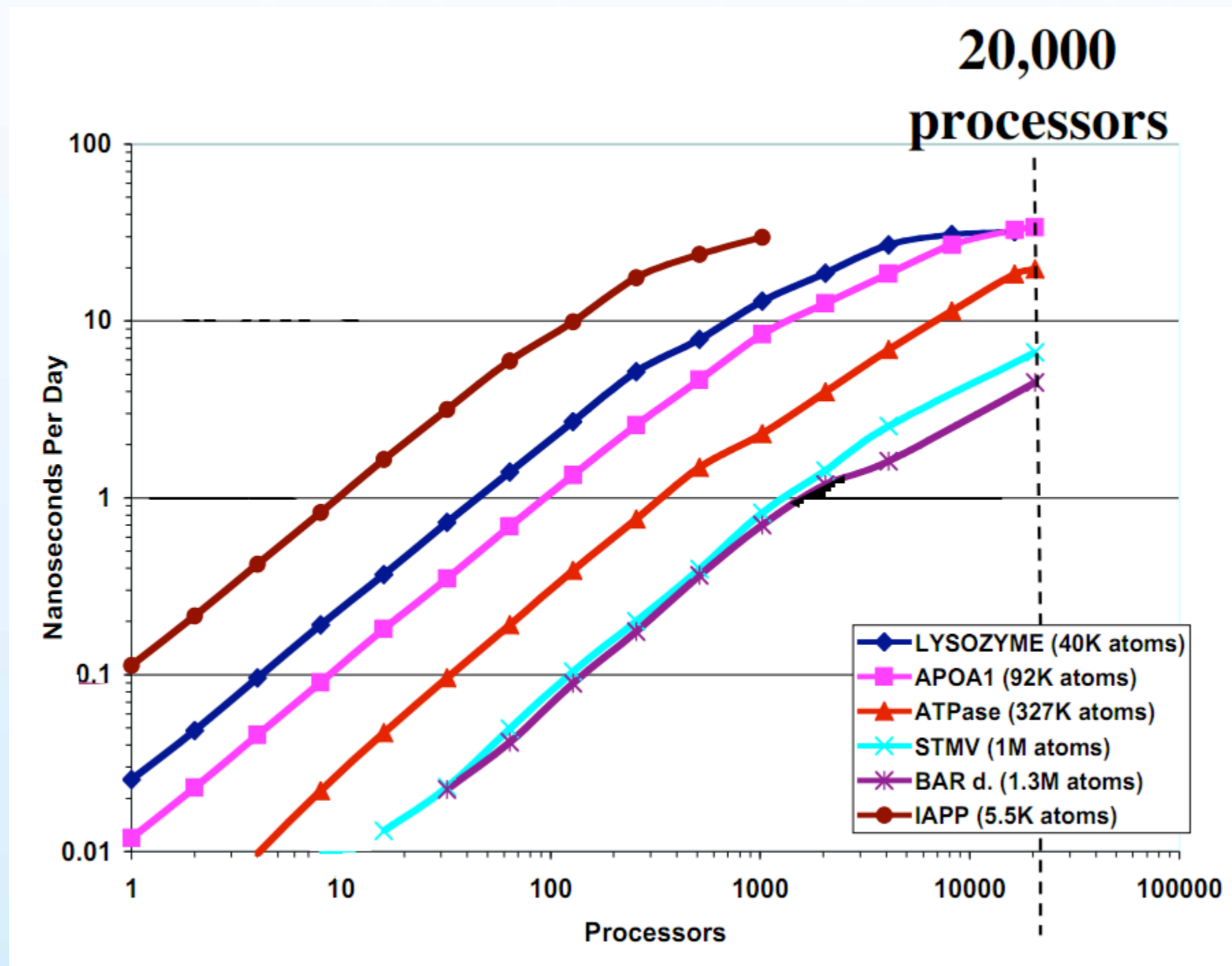
supercomputing (general, with GPU, ASIC, etc)



Domain decomposition

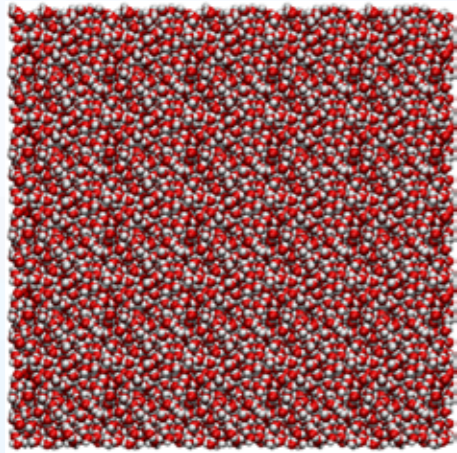


Parallel computing

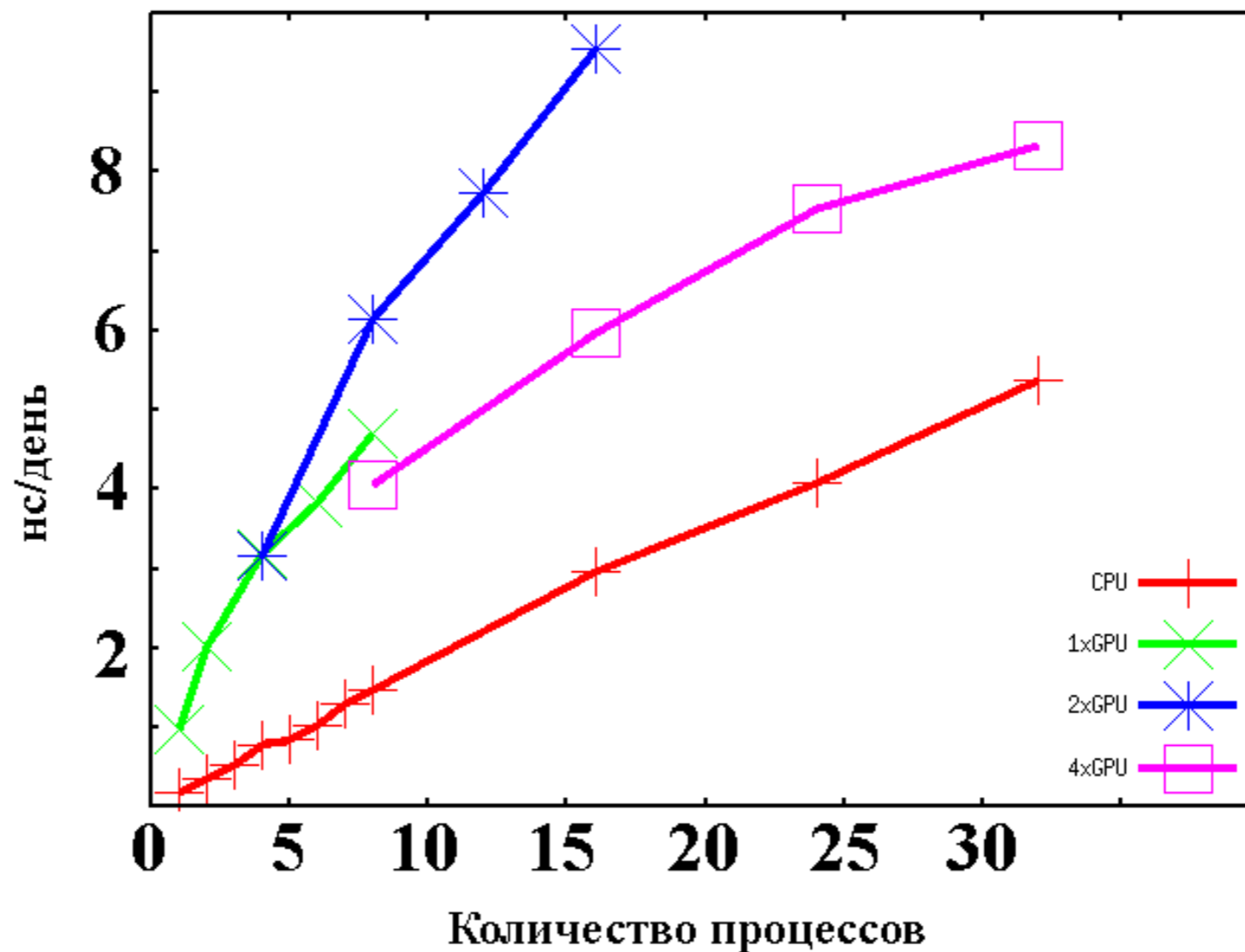
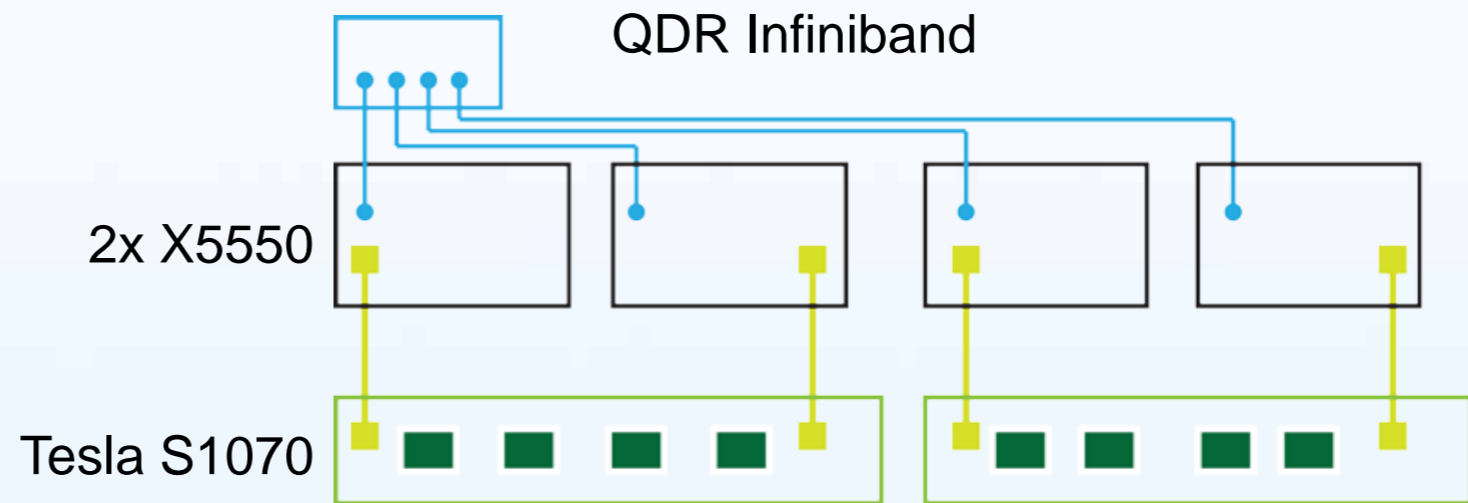


NMAD scaling (from the NAMMD website)

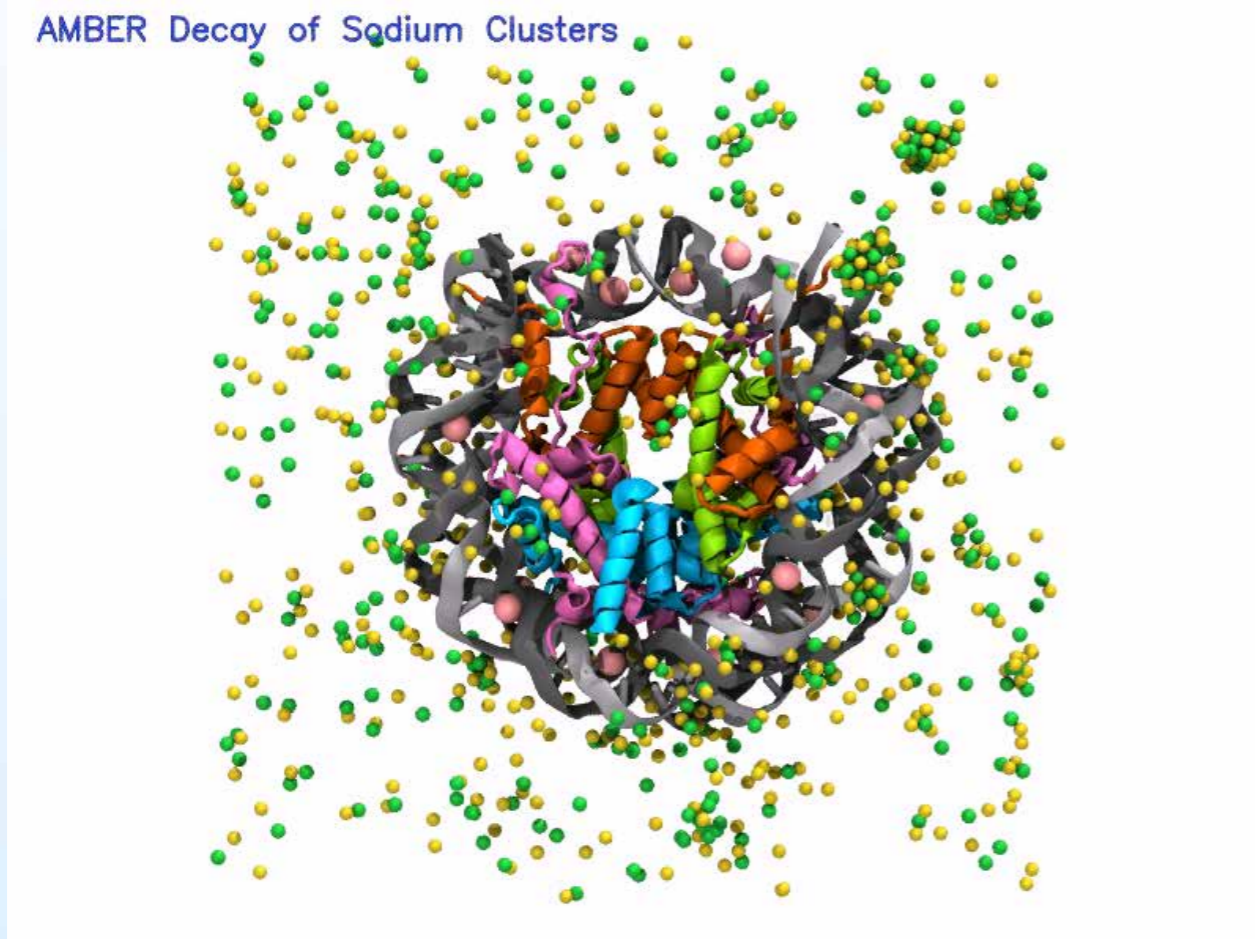
Scaling example



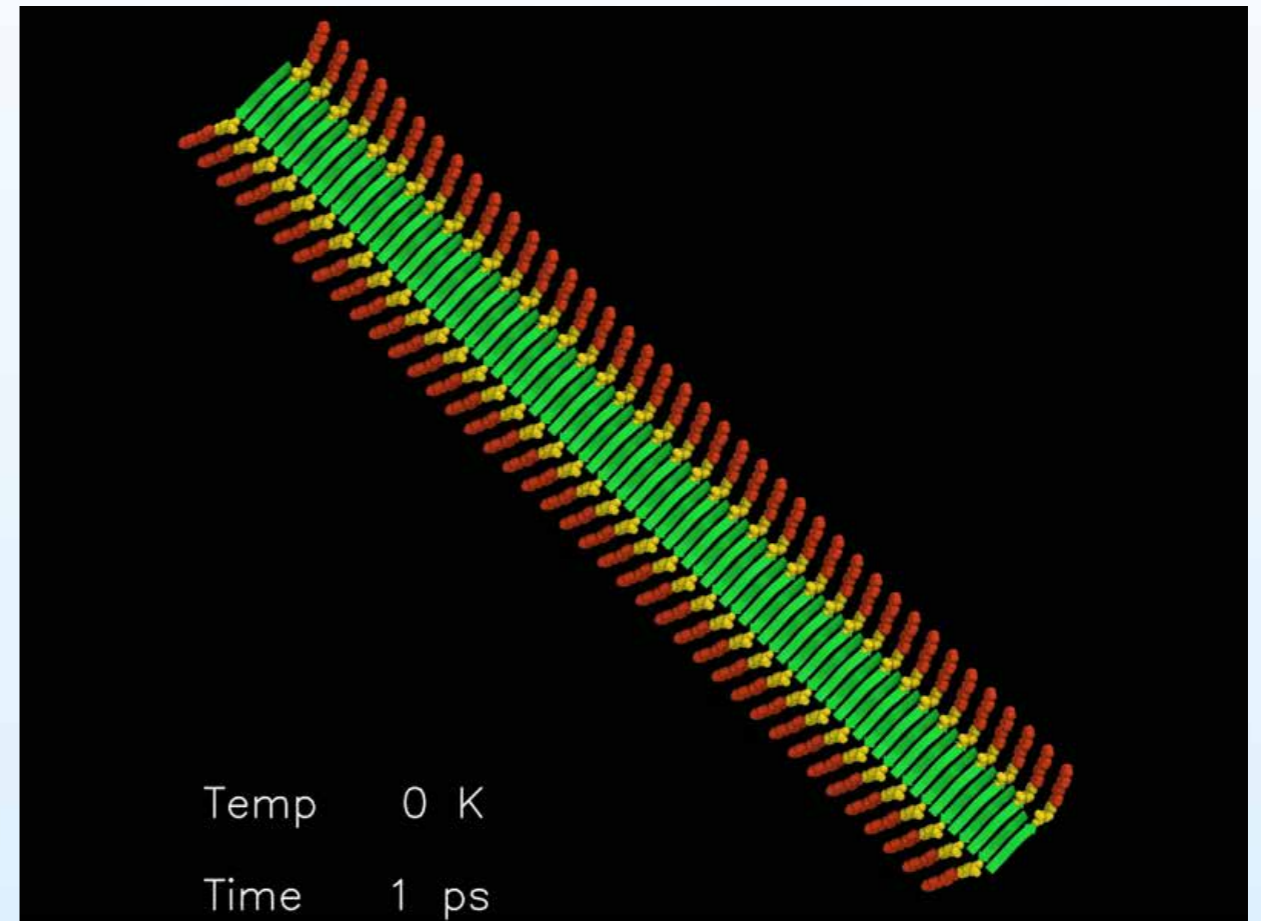
100K atoms
32K waters



Time and scale examples

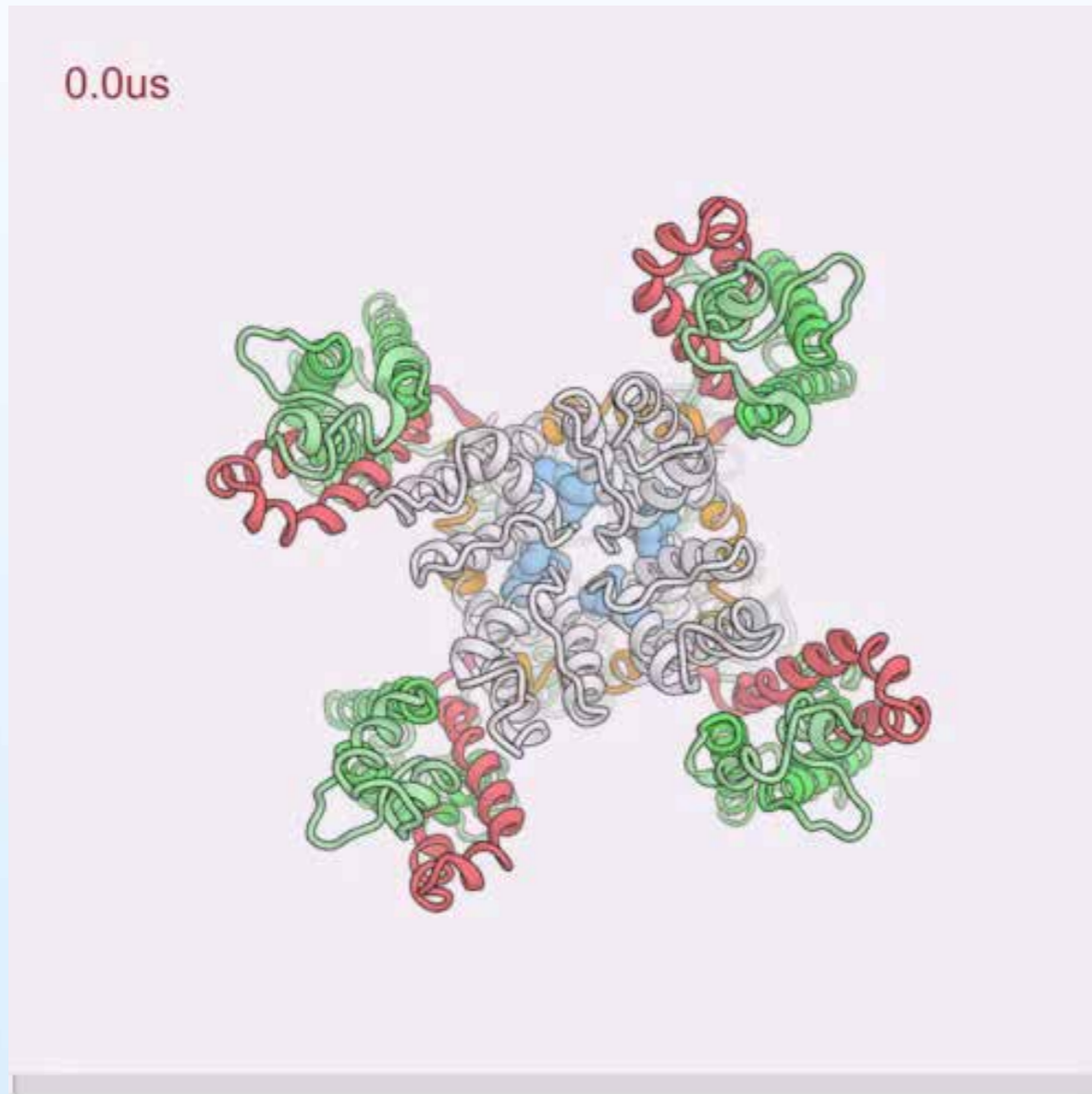


DNA-protein complex in
salt solution.
200 000 atoms



40 nm hybrid nanowire
from thiophenes and
peptides.
50 000 atoms

Time and scale examples

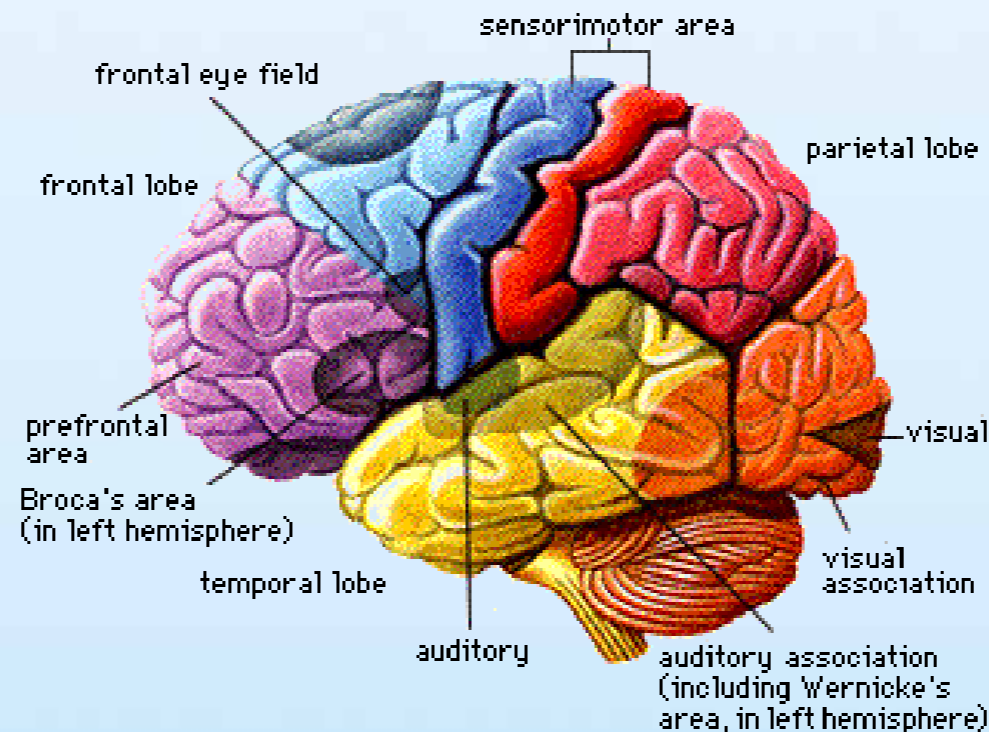


Potassium
voltage gated
channel
~250 μ s
~200 000 atoms

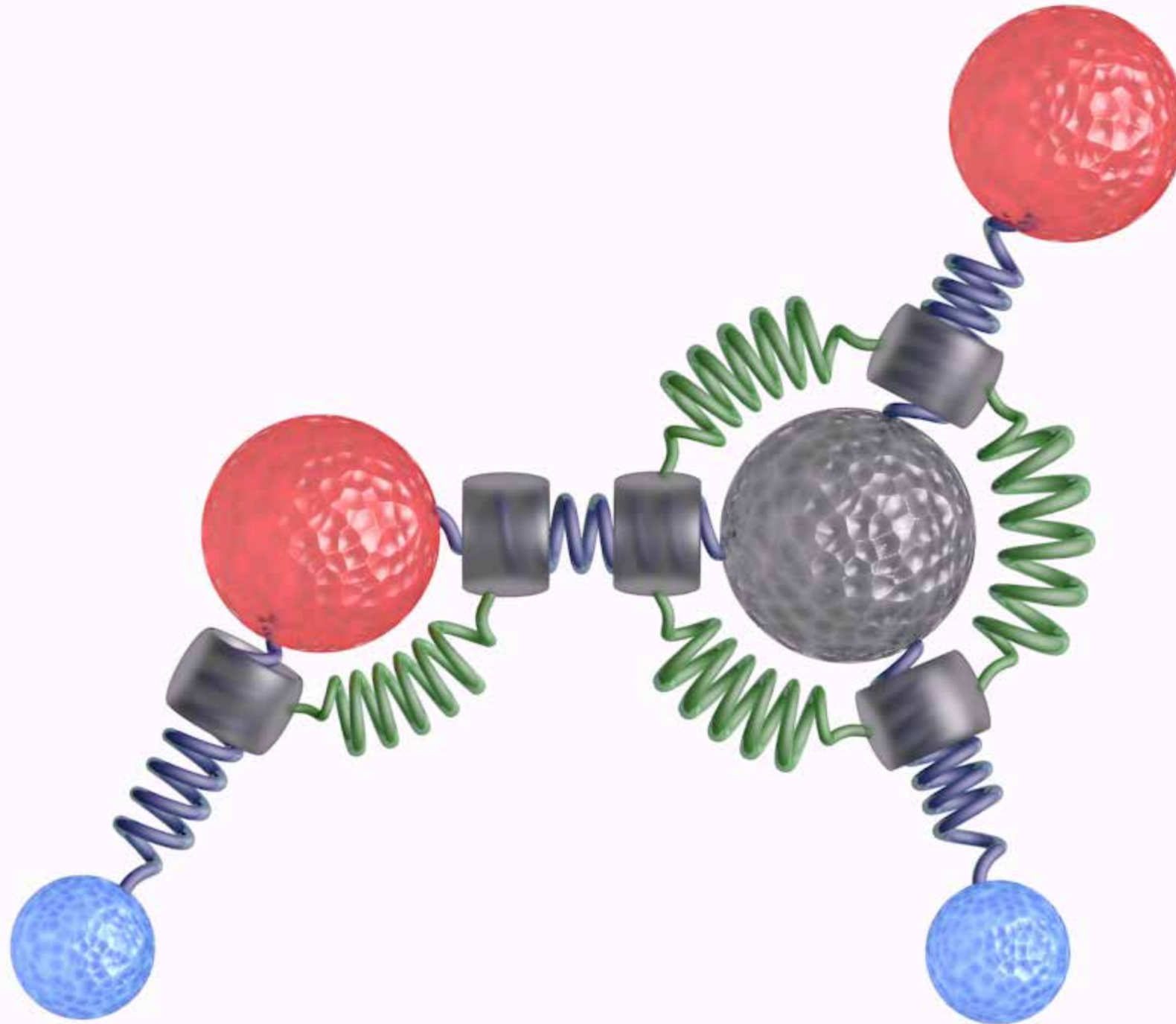
Science 2012, Morten Ø. Jensen et al. “Mechanism of Voltage Gating in Potassium Channels”

Advanced approaches

- Multiscale modeling: coarse graining, QM/MM etc.
- Advanced sampling: parallel tempering, meta dynamics, adaptive algorithms
- Thermodynamics: free energy profiles, reaction coordinates, alchemical transformations, etc



Thanks for attention!



<http://molsim.org/>

<http://www.youtube.com/molsimmsu>