



ZOLLVEREIN UNESCO WORLD
HERITAGE SITE
THE HEART OF THE RUHR AREA

Atomistic Computer Simulation of Hard and Soft Matter

Eckhard Spohr, Stéphane Kenmoe, Faculty for Chemistry
University Duisburg-Essen, Germany
eckhard.spohr@uni-due.de

8th African School of Physics
Marrakech, July 16, 2024



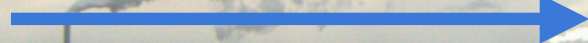
**ZOLLVEREIN UNESCO WORLD
HERITAGE SITE**
THE HEART OF THE RUHR AREA



- Dr. Stéphane Kenmoe
- Cameroonian
- Science Communication
- CASESMA (since 2019)
(Central African School of
Electronic Structure of
Materials)



Physics



Chemistry

Standard Model
Gravitation

**Electrodynamics
(Quantum Mechanics)**

The rest:

- too weak
- too strong



Atomistic Simulation?

- 1 • modeling of real physical (chemical / biological) systems
- 5 • a subdiscipline of statistical mechanics
- 3 • calculates thermodynamic properties
 - follows the 'real' trajectory of a system
 - study structure, evolution thereof
- 2 • Molecular Dynamics: studies the dynamics 6
 - individual atoms
- 4 • molecules
 - clusters
 - structural elements
 -
 - parts of living cells
 - materials
- 7 • chemical reaction dynamics

Atomistic Simulation?



- **modeling** of real physical (chemical / biological) systems
- a subdiscipline of **statistical mechanics**
- calculates **thermodynamic** properties
- follows the ‘real’ trajectory of a system
- studies structure, evolution thereof
- **Molecular Dynamics**: studies the dynamics
 - individual atoms
 - **molecules**
 - clusters
 - structural elements
 -
 - parts of living cells
 - **materials**
- chemical **reaction dynamics**

Modeling: Hamiltonians

- assume that we can write down a Hamiltonian of a system of atoms (usually assumed to be classical)
 - with or
 - without electronic degrees of freedom
- i.e., we often assume a first level of ‘coarse graining’
 - we treat the atom cores as the moving entities
 - electrons follow (adiabatically?)
 - **Born-Oppenheimer** approximation:
- electronic degrees of freedom:
 - quantum mechanically: various approximations to the **Schrödinger equation**
 - most frequently: Density functional Theory (**DFT**)
 - ...
 - empirical **force field**



Modeling

several 'directions' along which one can improve/
simplify Hamiltonian models:

- system size: as large as possible, as small as necessary
 - 10 atoms — 10^8 atoms
- simulation time: long enough to observe the relevant processes
 - 1 ps — 10 ps — 1 ns — ... — ms
- realism / simplification of the Hamiltonian:
 - is electronic structure important?
 - ground state properties only?
 - does every atom count? 'coarse graining'

Atomistic Simulation?

- **modeling** of real physical (chemical / biological) systems
- a subdiscipline of **statistical mechanics**
- calculates **thermodynamic** properties
- follows the 'real' trajectory of a system
- **Monte Carlo:** studies structure, evolution thereof
- **Molecular Dynamics:** also studies the dynamics
 - individual atoms
 - **molecules**
 - clusters
 - structural elements
 -
 - parts of living cells
 - **materials**
- chemical **reaction dynamics**



Monte Carlo

- principle: calculate multi-dimensional integrals

$$\langle \mathcal{A} \rangle_{NVT} = \int d\Gamma \rho_{NVT}(\Gamma) \mathcal{A}(\Gamma) \quad \langle \mathcal{A} \rangle_{NVT} = \left\langle \frac{\mathcal{A} \rho_{NVT}}{\rho} \right\rangle \quad \rho = \rho_{NVT}$$

- choose:

$$\rho_{NVT}(\Gamma) = \frac{1}{Z_{NVT}} \exp[-\beta \mathcal{V}(\Gamma)]$$

$$\frac{\rho_n}{\rho_m} = \frac{\exp(-\beta \mathcal{V}_n)}{\exp(-\beta \mathcal{V}_m)} = \exp[-\beta(\mathcal{V}_n - \mathcal{V}_m)]$$

$$\langle \mathcal{A} \rangle_{NVT} = \langle \mathcal{A} \rangle_{\text{Versuche}}$$

- algorithm :

$$\Pi_{mn} = \alpha_{mn} \quad \text{wenn } \rho_n \geq \rho_m \quad \text{und } m \neq n$$

$$\Pi_{mn} = \alpha_{mn}(\rho_n / \rho_m) \quad \text{wenn } \rho_n < \rho_m \quad \text{und } m \neq n \quad (\text{also } \Pi_{mn} < \alpha_{mn})$$

$$\Pi_{mm} = 1 - \sum_{n \neq m} \Pi_{mn}$$

$$\rho_m \Pi_{mn} = \rho_n \Pi_{nm} \quad \iff \quad \frac{\rho_n}{\rho_m} = \frac{\Pi_{mn}}{\Pi_{nm}}$$

Molecular Dynamics

- solve Newton's equations of motions for $3N$ dofs

$$q_i(t + \delta t) = q_i(t) + \dot{q}_i(t) \cdot \delta t + \frac{1}{2}\ddot{q}_i(t) \cdot (\delta t)^2 \pm \dots$$

$$q_i(t - \delta t) = q_i(t) + \dot{q}_i(t) \cdot (-\delta t) + \frac{1}{2}\ddot{q}_i(t) \cdot (-\delta t)^2 \pm \dots$$

- add up: $q_i(t + \delta t) = 2q_i(t) - q_i(t - \delta t) + \ddot{q}_i(t) \cdot (\delta t)^2$

- generates configurations with equal a priori probability,
- i.e. a **microcanonical NVE ensemble**
- averages are simple sums divided by # of configurations

used already in first simulations in 1960s

A Simple MD Code

```
nstepmax = ????
```

```
step = 0
```

```
model = your_choice_of_hamiltonian
```

```
read_starting_configuration()
```

```
while not happy or step < nstepmax:
```

```
    propagate_coordinates_a_bit(delta_t)
```

```
    calculate_forces(model)
```

```
    collect_some_data_sometimes(your_question)
```

```
    write_conf_to_disk_every_now_and_then()
```

```
    happy = your_decision_based_on...()
```

```
    step += 1
```

```
write_final_configuration()
```


Open Source MD Codes

- Gromacs ($> 50,000$ lines)
- Amber
- Namd
- OpenMM
- Lammmps ($> 640,000$ lines)
- CP2K ($> 1,150,000$ lines)
-

Open Source MD Codes

what makes these codes so large?

- support for different interaction models
- support for electronic structure calculations
- support for handling hundreds of atoms
- support for handling millions of atoms
- support for building geometric models of chemical systems
- I/O
- optimization
- on-the-fly analysis
- support for various ensembles / boundary conditions
- ... 'infrastructure'

Open Source MD Codes

as of 2024:


- infrastructure well developed
- no (?) need to write own codes
- learn how to
 - ask a good scientific question
 - select a suitable code
 - use it
 - understand its limitations

What do you want to know?

- what is the physical/chemical process of interest?
- is the system ordered or disordered?
- spatial structure: assume it or produce it?
- relevant time scale of process?
- relevant time scale of relaxation to equilibrium?
- do you want to study a system
 - in equilibrium?
 - out of equilibrium?
- dominant interactions?

... choose simulation code / system size / interaction model / simulation length /

Atomistic Simulation?

- **modeling** of real physical (chemical / biological) systems
- a subdiscipline of **statistical mechanics**
-  calculates **thermodynamic** properties
- follows the 'real' trajectory of a system
- studies structure, evolution thereof
- **Molecular Dynamics**: studies the dynamics of
 - individual atoms
 - **molecules**
 - clusters
 - structural elements
 -
 - parts of living cells
 - **materials**
- chemical **reaction dynamics**

Thermostats / Barostats

- Nosé-Hoover thermostat: NVT ensemble

$$H = \sum_{i=1}^N \frac{\mathbf{p}_i^2}{2m_i} + U(\mathbf{r}_1, \mathbf{r}_2, \dots, \mathbf{r}_N) + \frac{p_\xi^2}{2Q} + N_f kT \xi$$

- Parrinello-Rahman barostat: NPH ensemble

$$H = E_{\text{pot}} + E_{\text{kin}} + \sum_i P_{ii} V + \sum_{i,j} \frac{1}{2} W_{ij} \left(\frac{db_{ij}}{dt} \right)^2$$

- combination: NPT ensemble

extended system Hamiltonians

Thermodynamics: Eq. of State

- pressure, volume, temperature are also simple averages
- establish a simulation procedure that allows
 - calculation of $\langle V \rangle(T, P)$
 - or $\langle P \rangle(T, V)$
- \implies equation of state
 - very important in technical thermodynamics
 - also for matter under extreme conditions

Atomistic Simulation?

- **modeling** of real physical (chemical / biological) systems
- a subdiscipline of **statistical mechanics**
- calculates **thermodynamic** properties
- follows the ‘real’ trajectory of a system
- studies structure, evolution thereof
- **Molecular Dynamics**: studies the dynamics of
 - individual atoms
 - **molecules**
 - clusters
 - structural elements
 -
 - parts of living cells
 - **materials**
- chemical **reaction dynamics**



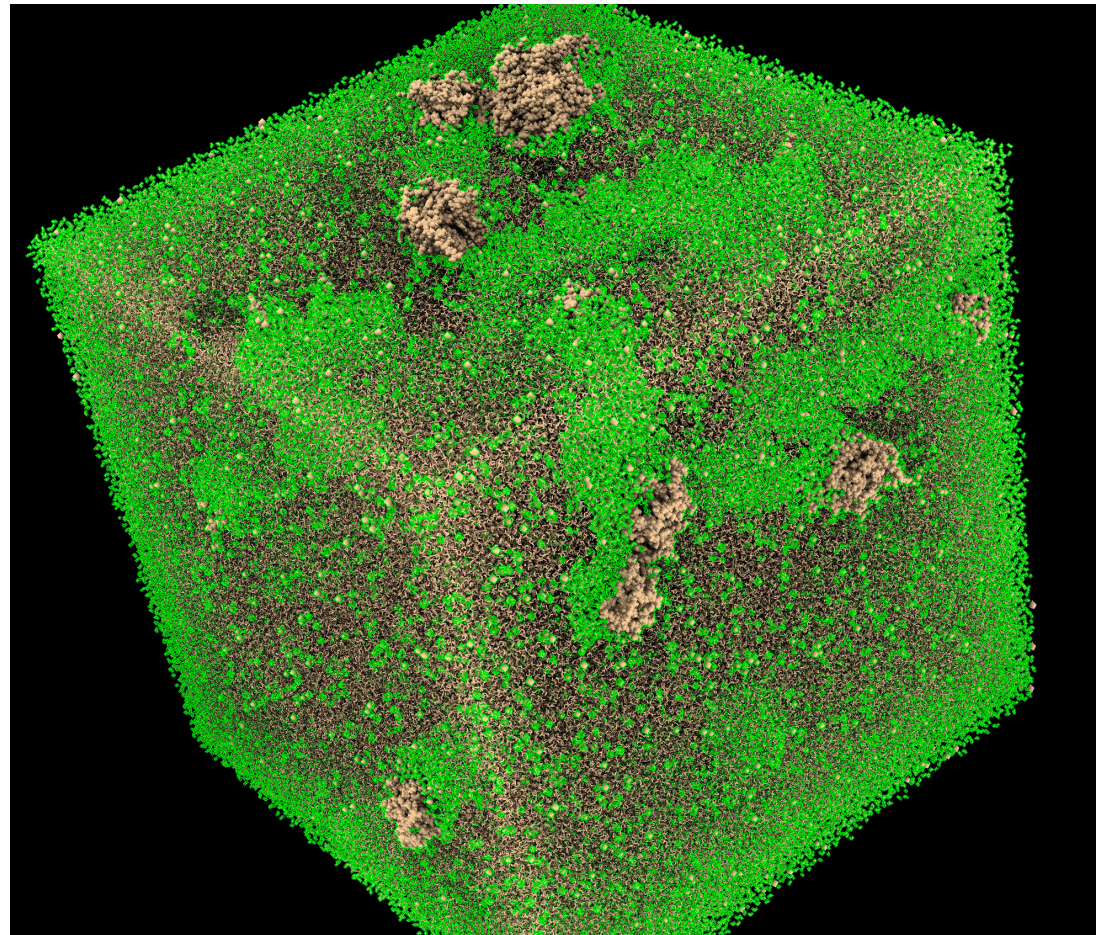
A Simple Protein Force Field

$$\begin{aligned} V(\{r\}) = & \\ & \sum_{\text{bonds}} \frac{k_i}{2} (r_i - r_{eq,i})^2 \\ + & \sum_{\text{angles}} \frac{k_i}{2} (\theta_i - \theta_{eq,i})^2 \\ + & \sum_{\text{torsions}} \frac{v_i}{2} [1 + \cos(n_i \omega_i - \gamma_i)] \\ + & \sum_{\text{pairs } i-j} 4\epsilon_{ij} \left[\left(\frac{\sigma_{ij}}{r_{ij}} \right)^{12} - \left(\frac{\sigma_{ij}}{r_{ij}} \right)^6 \right] \\ + & \sum_{\text{pairs } i-j} \frac{1}{4\pi\epsilon_0} \frac{q_i q_j}{r_{ij}} \end{aligned}$$

Simple Averages

A Model for a Protein Solution

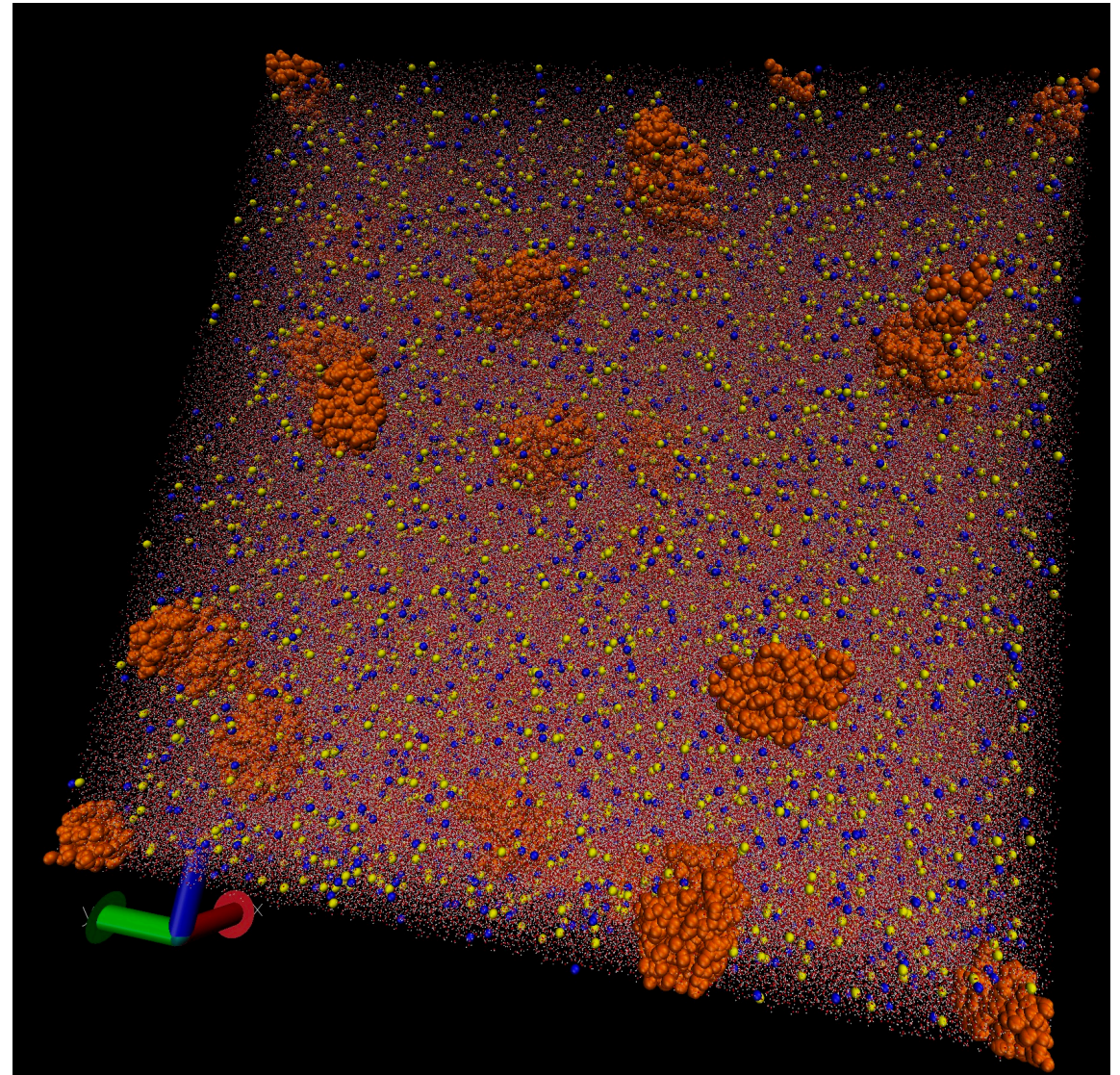
- solution:
 - pure water or
 - 0.25 mol/l NaCl in water
 - or any other solution of electrolyte
- 64 HEWL protein molecules
- altogether 3.4 million atoms
- running Gromacs with
 - Gromacs force field
 - Amber force field
- vary the nature of the electrolyte



Simple Averages

A Model for a Protein Solution

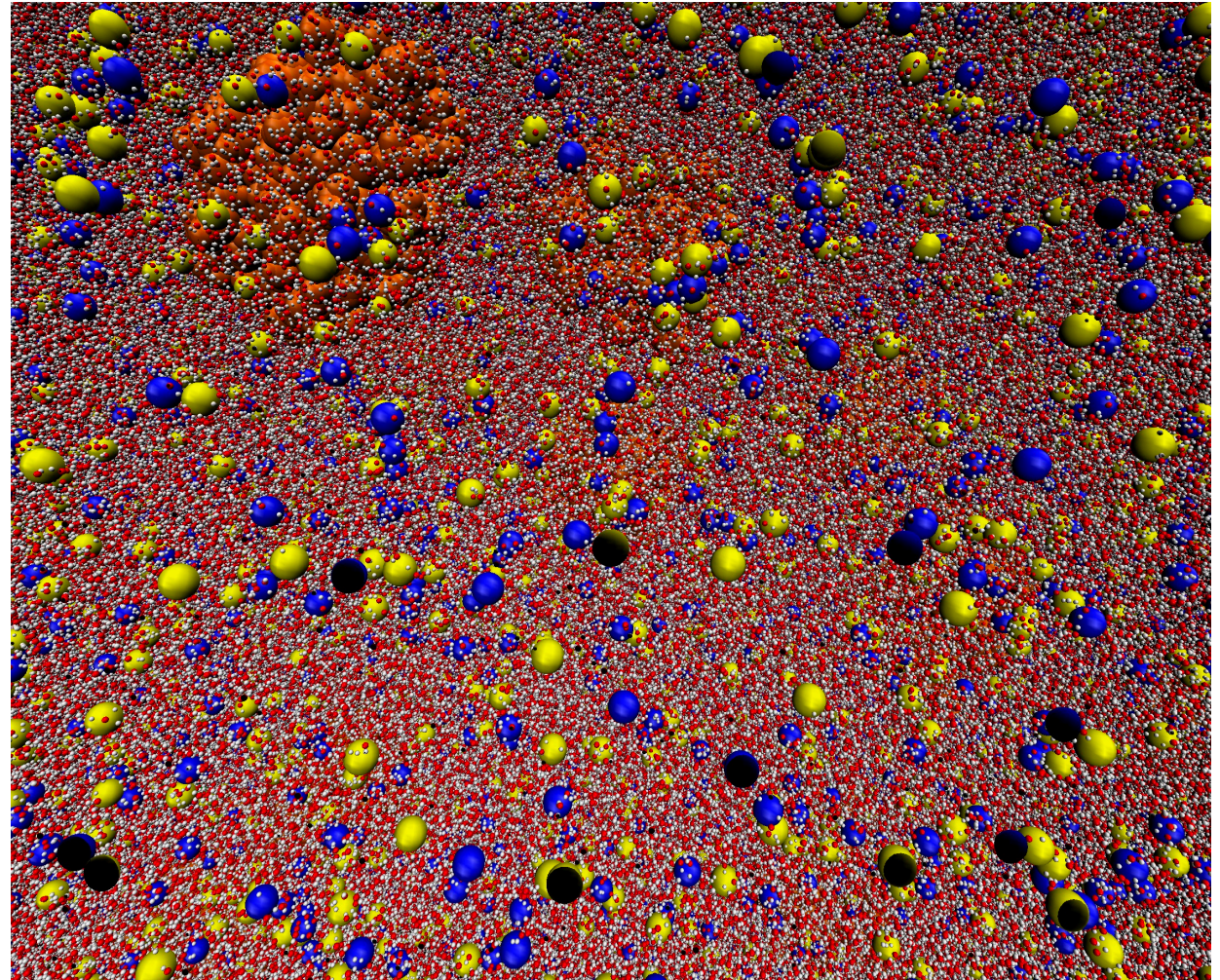
- solution:
 - pure water or
 - 0.25 mol/l NaCl in water
 - or any other solution of electrolyte
- 64 protein molecules
- altogether 3.4 million atoms
- running Gromacs with
 - Gromacs force field
 - Amber force field
- vary the nature of the electrolyte



Simple Averages

A Model for a Protein Solution

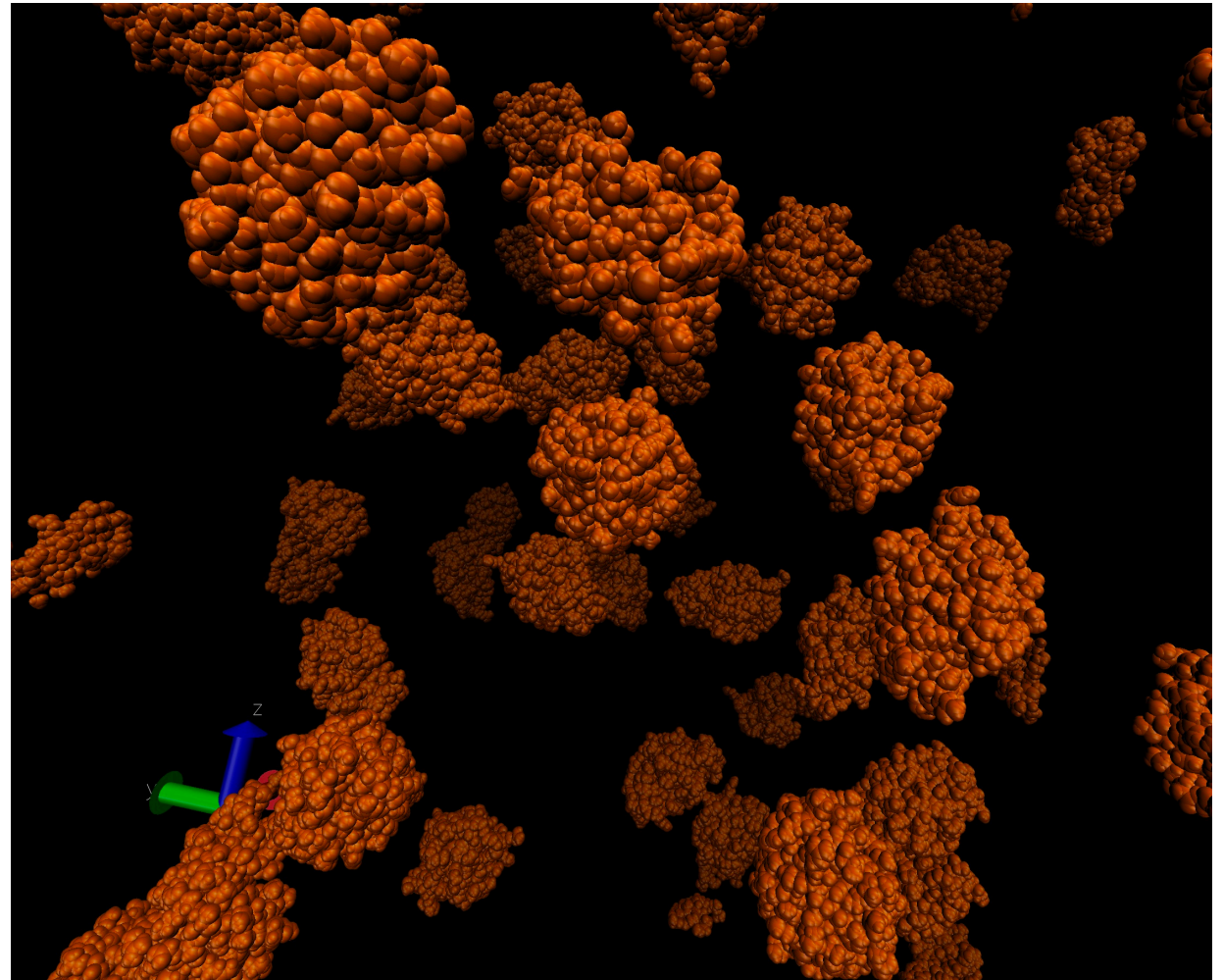
- solution:
 - pure water or
 - 0.25 mol/l NaCl in water
 - or any other solution of electrolyte
- 64 protein molecules
- altogether 3.4 million atoms
- running Gromacs with
 - Gromacs force field
 - Amber force field
- vary the nature of the electrolyte



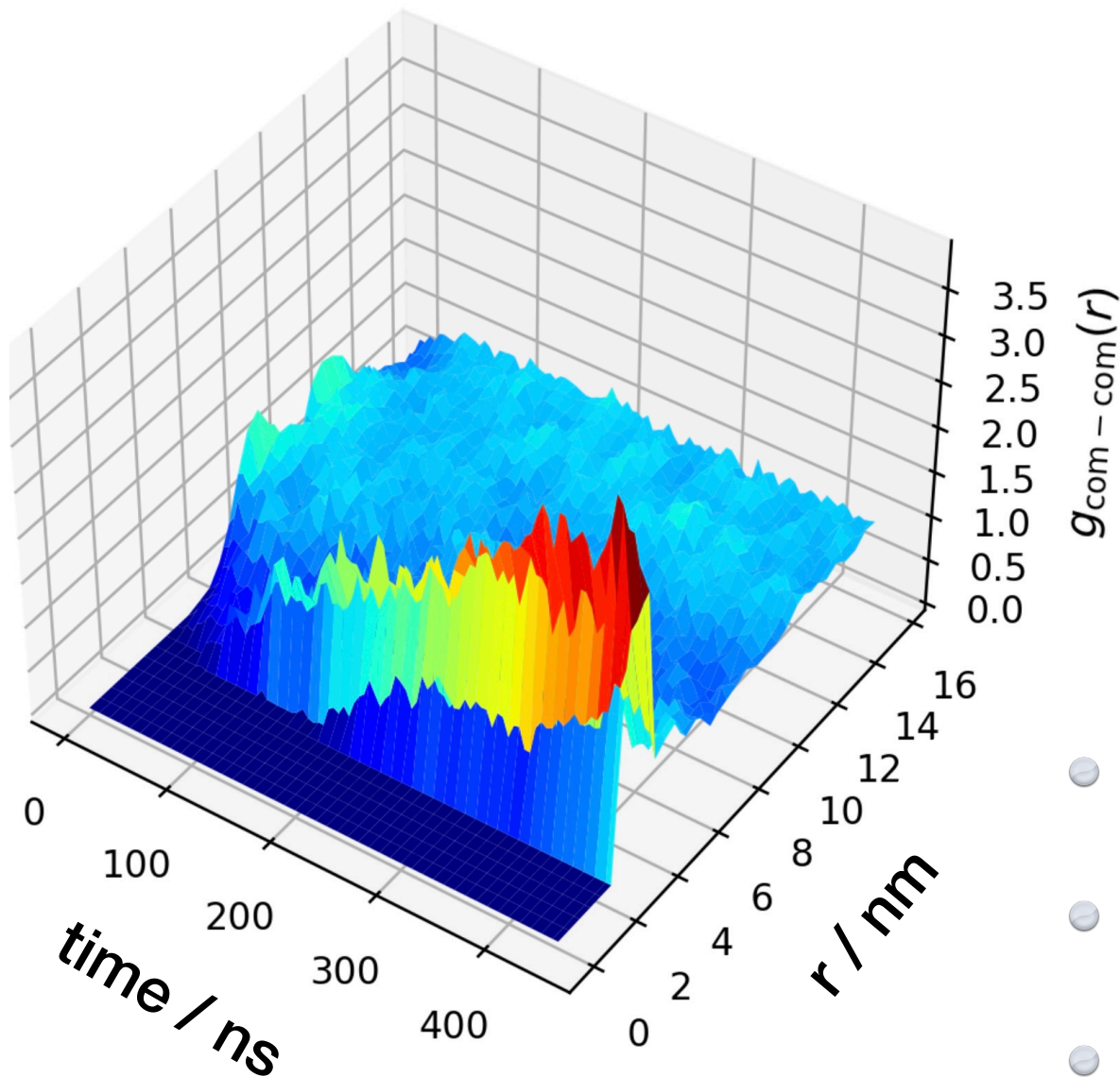
Simple Averages

A Model for a Protein Solution

- solution:
 - pure water or
 - 0.25 mol/l NaCl in water
 - or any other solution of electrolyte
- 64 protein molecules
- altogether 3.4 million atoms
- running Gromacs with
 - Gromacs force field
 - Amber force field
- vary the nature of the electrolyte



Simple Averages



radial distribution function

equilibration:
nuisance or objective?

- Simulation was started from cubic 4x4x4 lattice of HEWL molecules
- Cubic structure vanishes over 50 ns
- Slow buildup of nearest neighbour structure over 400+ ns

Atomistic Simulation?

5

- **modeling** of real physical (chemical / biological) systems
- a subdiscipline of **statistical mechanics**
- calculates **thermodynamic** properties
- follows the ‘real’ trajectory of a system
- studies structure, evolution thereof
- **Molecular Dynamics**: studies the dynamics of
 - individual atoms
 - **molecules**
 - clusters
 - structural elements
 -
 - parts of living cells
 - **materials**
- chemical **reaction dynamics**

Internal Energy

$$A = -kT \ln Q$$

$$U = A + TS$$

$$= \frac{\int d\Gamma E(\{q_k, p_k\}) \exp \left[-\frac{E(\{q_k, p_k\})}{kT} \right]}{\int d\Gamma \exp \left[-\frac{E(\{q_k, p_k\})}{kT} \right]}$$

$$= \frac{\int d\Gamma (\mathcal{T}(\{p_k\}) + \mathcal{V}(\{q_k\})) \exp \left[-\frac{\mathcal{T}(\{p_k\}) + \mathcal{V}(\{q_k\})}{kT} \right]}{\int d\Gamma \exp \left[-\frac{\mathcal{T}(\{p_k\}) + \mathcal{V}(\{q_k\})}{kT} \right]}$$

$$= \langle \mathcal{T} \rangle + \frac{\int dq_1 dq_2 \dots dq_{3N} \mathcal{V}(\{q_k\}) \exp \left[-\frac{\mathcal{V}(\{q_k\})}{kT} \right]}{Z_N}$$

$$= \langle \mathcal{T} \rangle + \langle \mathcal{V} \rangle = \frac{3N}{2} kT + \langle \mathcal{V} \rangle$$

internal energy is a simple average: a single simulation suffices

Free Energy

$$\begin{aligned}\Delta A &= A_Y - A_X = -kT \ln \frac{Q_Y}{Q_X} \\ &= -kT \ln \left\{ \frac{\int d\Gamma \exp \left[-\mathcal{H}_Y(\{q_k, p_k\}) / kT \right]}{\int d\Gamma \exp \left[-\mathcal{H}_X(\{q_k, p_k\}) / kT \right]} \right\} \\ &= -kT \ln \left\{ \frac{\int d\Gamma \exp \left[-\mathcal{H}_Y / kT \right] \exp \left[+\mathcal{H}_X / kT \right] \exp \left[-\mathcal{H}_X / kT \right]}{\int d\Gamma \exp \left[-\mathcal{H}_X(\{q_k, p_k\}) / kT \right]} \right\} \\ &= -kT \ln \left\{ \frac{\int d\Gamma \exp \left[-\mathcal{H}_Y / kT \right] \exp \left[+\mathcal{H}_X / kT \right] \exp \left[-\mathcal{H}_X / kT \right]}{\int d\Gamma \exp \left[-\mathcal{H}_X(\{q_k, p_k\}) / kT \right]} \right\} \\ &= -kT \ln \left\{ \int d\Gamma \exp \left[-\mathcal{H}_Y / kT \right] \exp \left[+\mathcal{H}_X / kT \right] \rho_X(\{q_k, p_k\}) \right\} \\ &= -kT \ln \langle \exp \left[-(\mathcal{H}_Y - \mathcal{H}_X) / kT \right] \rangle_X\end{aligned}$$

free energy: exponential average (Boltzmann factor):
difficult to converge

Free Energy: exponential averages

$$A = -kT \ln Q$$

$$= -kT \ln \int \int dq_1 dq_2 \dots dq_{3N} \int dp_1 dp_2 \dots dp_{3N} \exp \left[-\mathcal{H}(\{q_k, p_k\}) / kT \right]$$

$$\Delta A = A_Y - A_X$$

$$= -kT \ln \langle \exp \left[-(\mathcal{H}_Y - \mathcal{H}_X) / kT \right] \rangle_X$$

$$-\Delta A = A_X - A_Y$$

$$= -kT \ln \langle \exp \left[-(\mathcal{H}_X - \mathcal{H}_Y) / kT \right] \rangle_Y$$

====> a new 'discipline': computational alchemy

Transition State Theory

- important concept in physical chemistry
- basic assumption
 - A chemical reaction proceeds to completion once the
 - activation barrier (maximum of the minimum energy path)
- has been reached
- probability proportional to Boltzmann factor $e^{-\Delta E_a/kT}$
- if activation energy ΔE_a is large, getting there is a **rare event**
- ??? (get there stepwise, somehow) ???
- dynamic corrections: Transmission factor $\kappa < 1$

Free Energy Differences

- free energy differences between 2 chemical compounds, e.g. $X = \text{ethanol (C}_2\text{H}_5\text{OH)}$ and $Y = \text{thioethanol (C}_2\text{H}_5\text{SH)}$
 - Problem:
 - distribution of states X and Y in phase space do NOT overlap (S is 'larger' than 'O')
 - ====> bad (inefficient) sampling
- This problem typically occurs when for a given phase space point $|H_X - H_Y| \gg kT$
- problem solution: simulate intermediate steps

Free Energy Differences

$$X_1 = \text{ethanol}$$

$$Y_1 = 0.9 * \text{Ethanol} + 0.1 * \text{Ethanthiol}$$

$$X_2 = 0.9 * \text{Ethanol} + 0.1 * \text{Ethanthiol}$$

$$Y_2 = 0.8 * \text{Ethanol} + 0.2 * \text{Ethanthiol}$$

$$X_3 = 0.8 * \text{Ethanol} + 0.2 * \text{Ethanthiol}$$

$$Y_3 = 0.7 * \text{Ethanol} + 0.3 * \text{Ethanthiol}$$

...

$$X_{11} = 0.1 * \text{Ethanol} + 0.9 * \text{Ethanthiol}$$

$$Y_{11} = \text{Ethanthiol}$$

$$\begin{aligned} \Delta A &= A(Y_1) - A(X_1) + A(Y_2) - A(X_2) + \dots \\ &= A(Y_{11}) - A(X_1) \end{aligned}$$

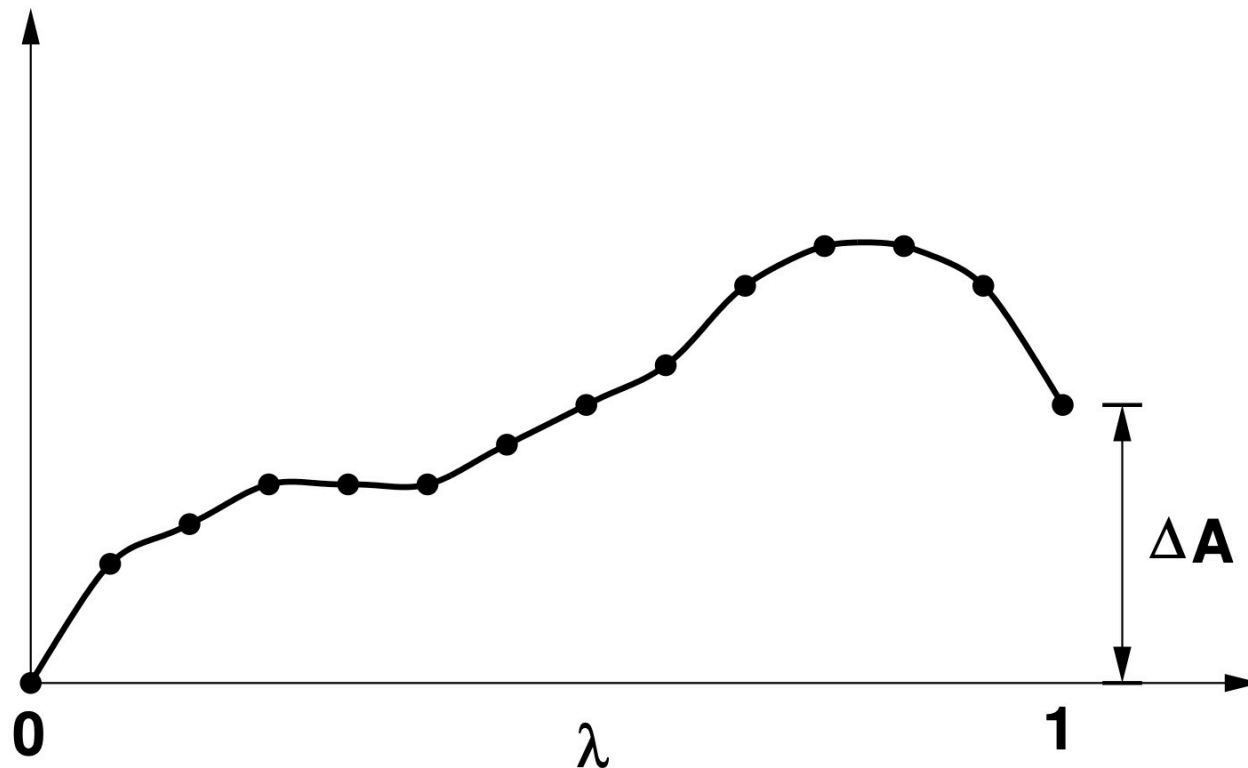
- generate unphysical (intermediate) ensembles
- need to interpolate Hamiltonian somehow

Free Energy Differences

$\Delta A(\lambda_i \rightarrow \lambda_{i+1})$ “forward sampling”

$\Delta A(\lambda_i \rightarrow \lambda_{i-1})$ “backward sampling”

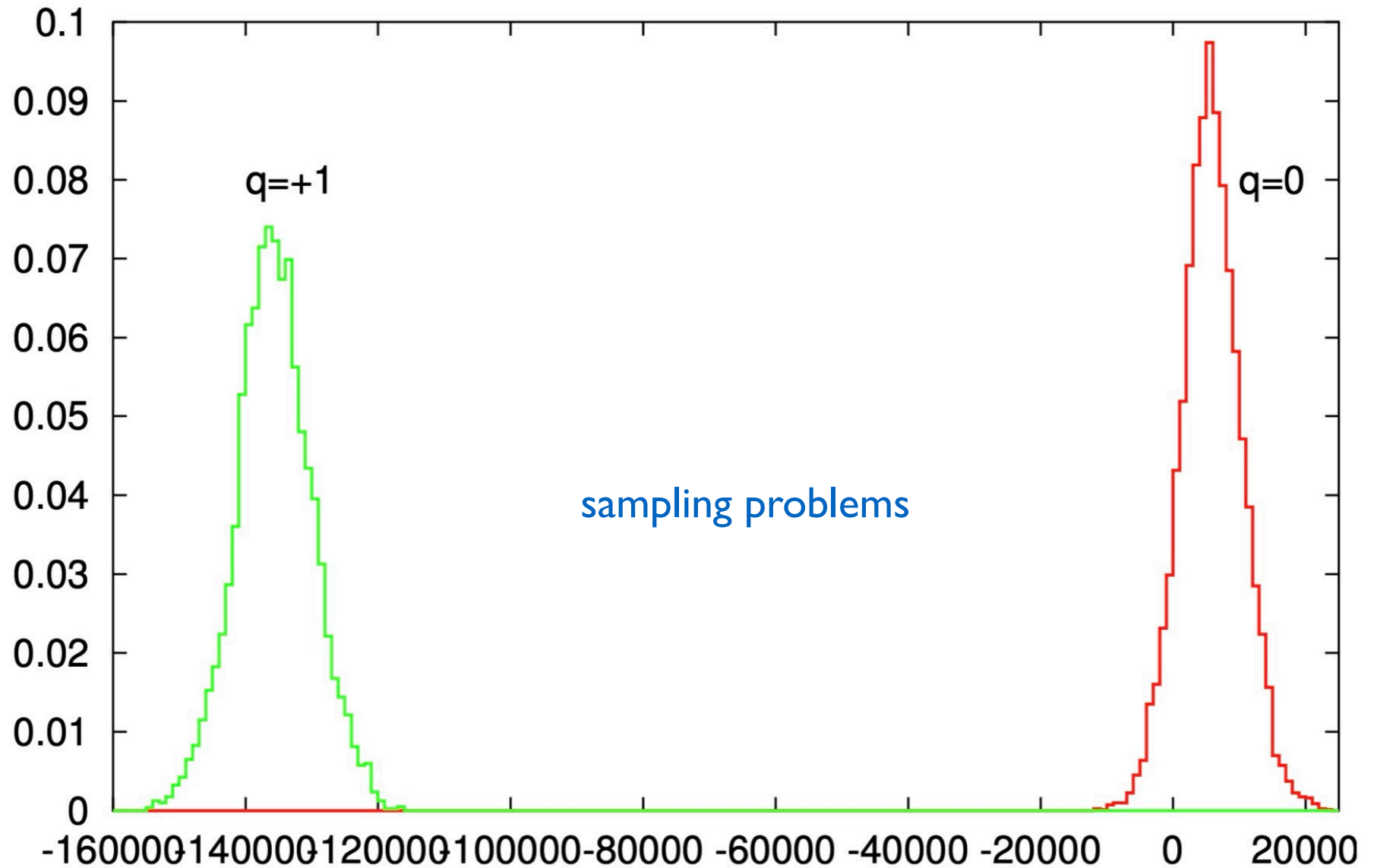
beide “double-wide sampling”



Example: Discharge of Na^+ (Born)

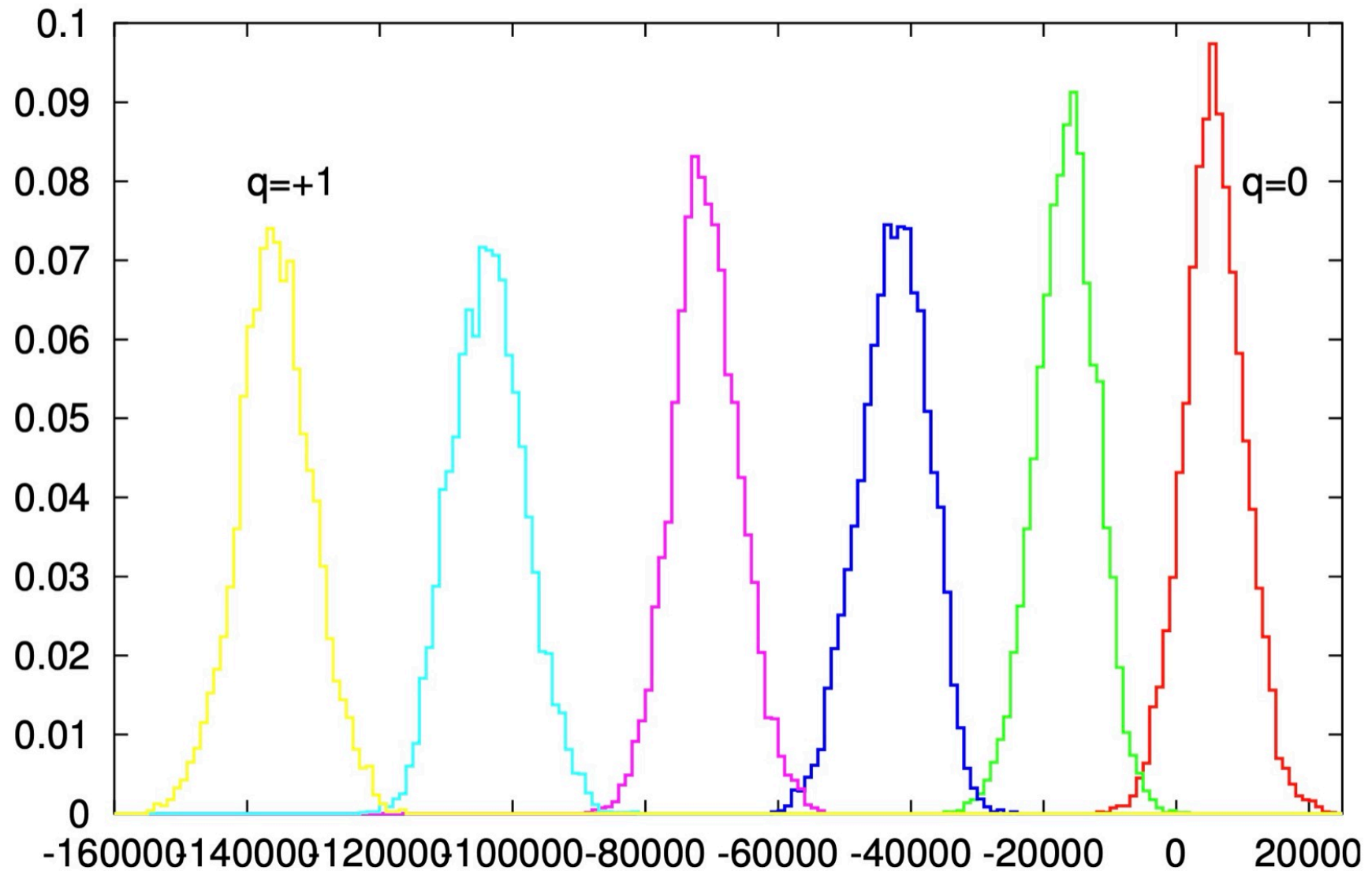
- generate II simulations where the sodium ion has charge $1e, 0.9e, 0.8e, \dots, 0.1e, 0e$
- in each simulation, water solvates the partially charged ion differently, creating a different electrostatic potential (and distribution)
- the potential distribution drives the charging/discharging of the ion (adiabatic principle)
- these II simulations can be analysed in 3 series:
 - $q = 0, 1$ $\Delta H = \phi$
 - $q = 0, 0.2, 0.4, 0.6, 0.8, 1.0$ $\Delta H = 0.2\phi$
 - $q = 0, 0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9, 1.0$ $\Delta H = 0.1\phi$
- $\Delta H \rightarrow kT$

Electrostatic Potential Histograms



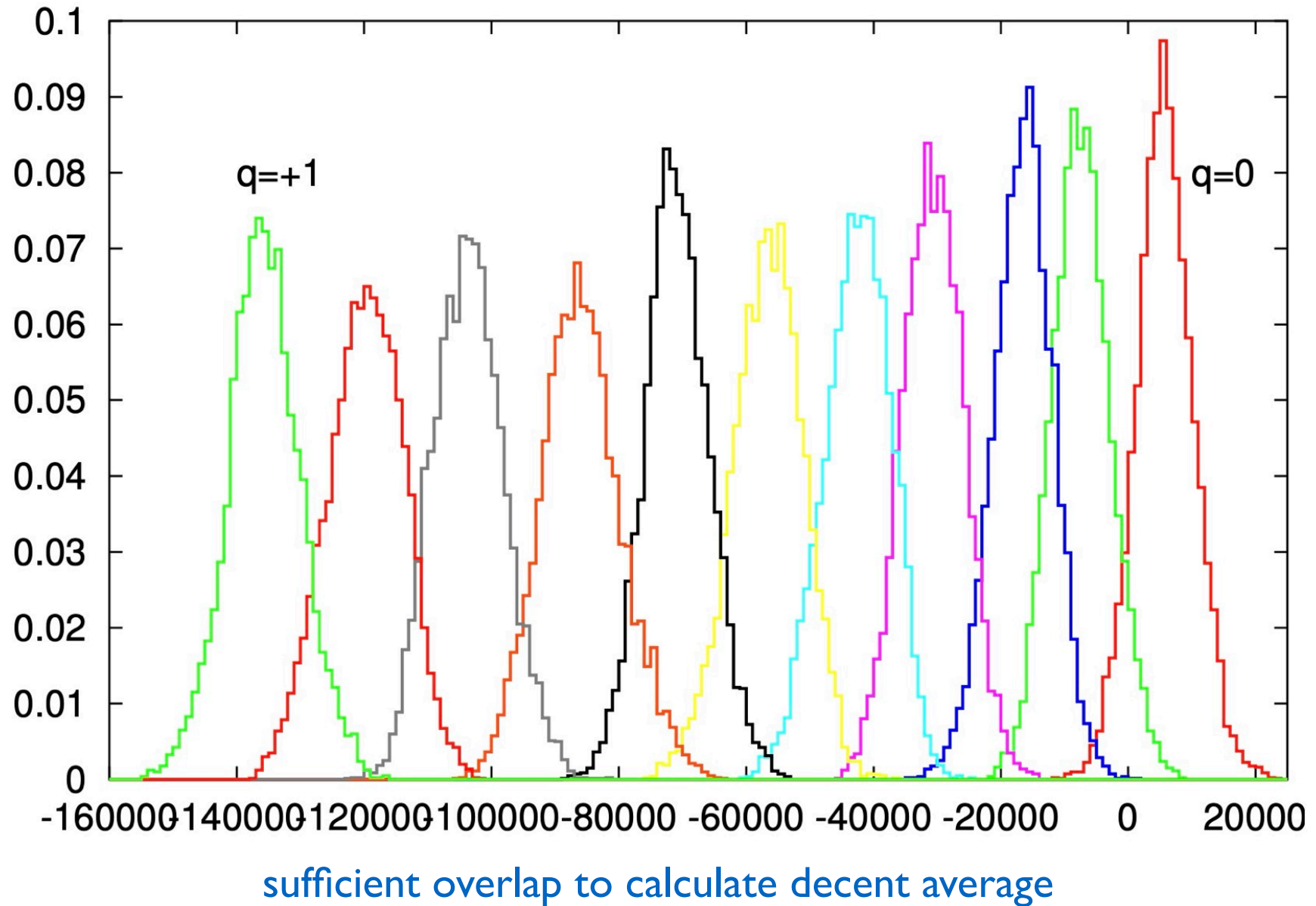
no overlap between initial and final configurations: impossible to calculate exp. average

Potential Histograms



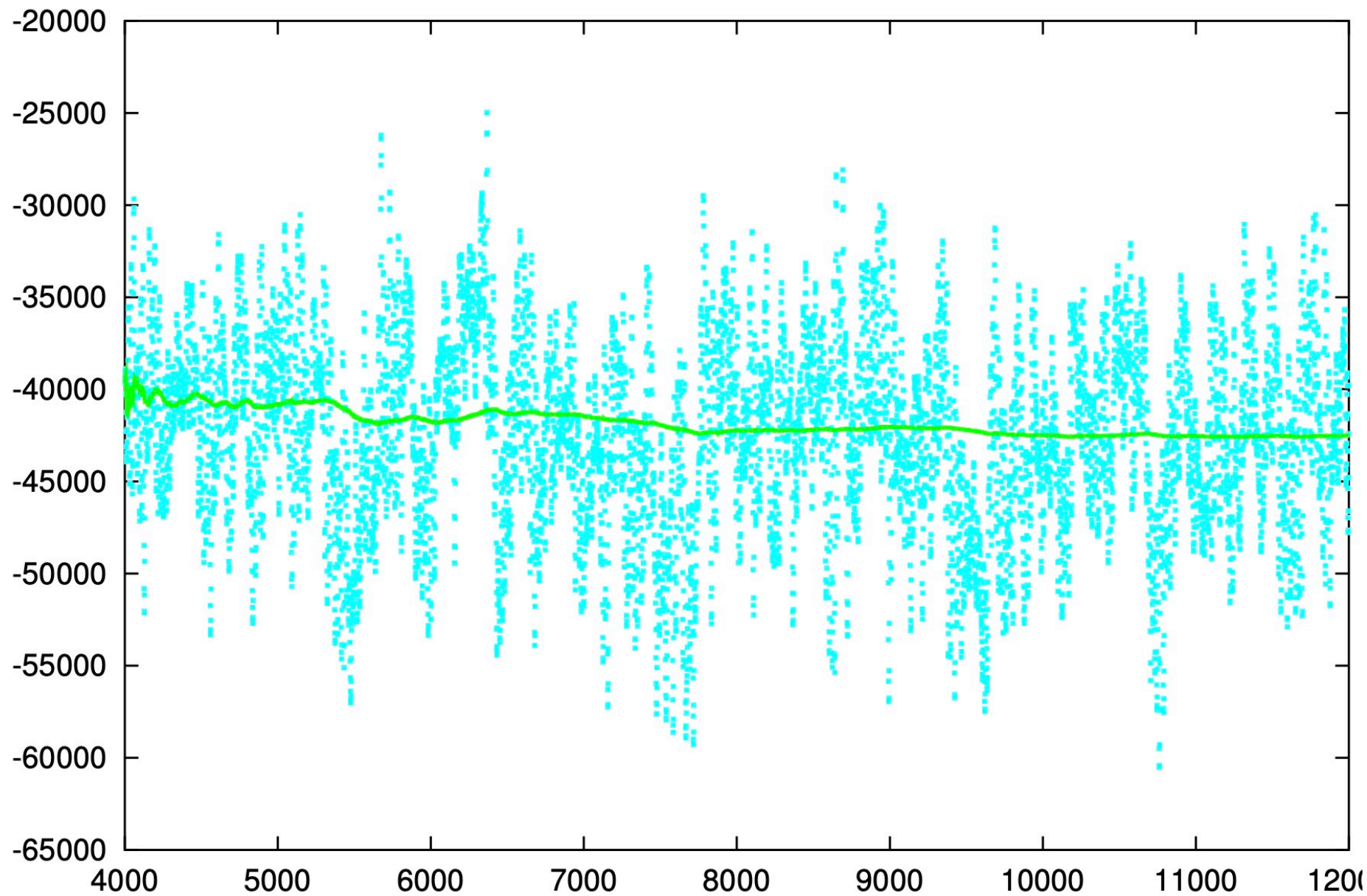
almost no overlap between initial and final configurations: average will be poor!

Potential Histograms



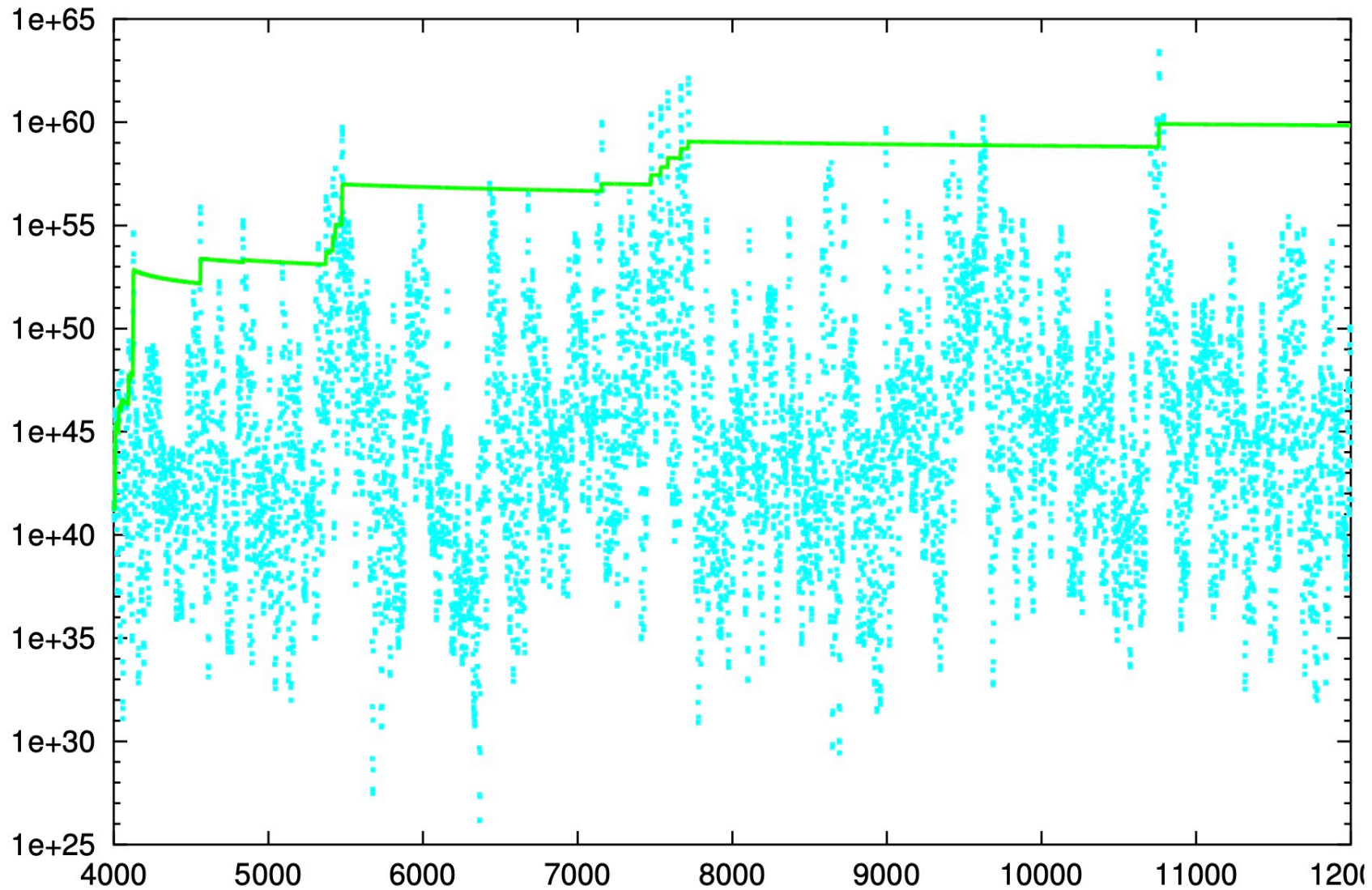
Running Averages

Delta H and running average



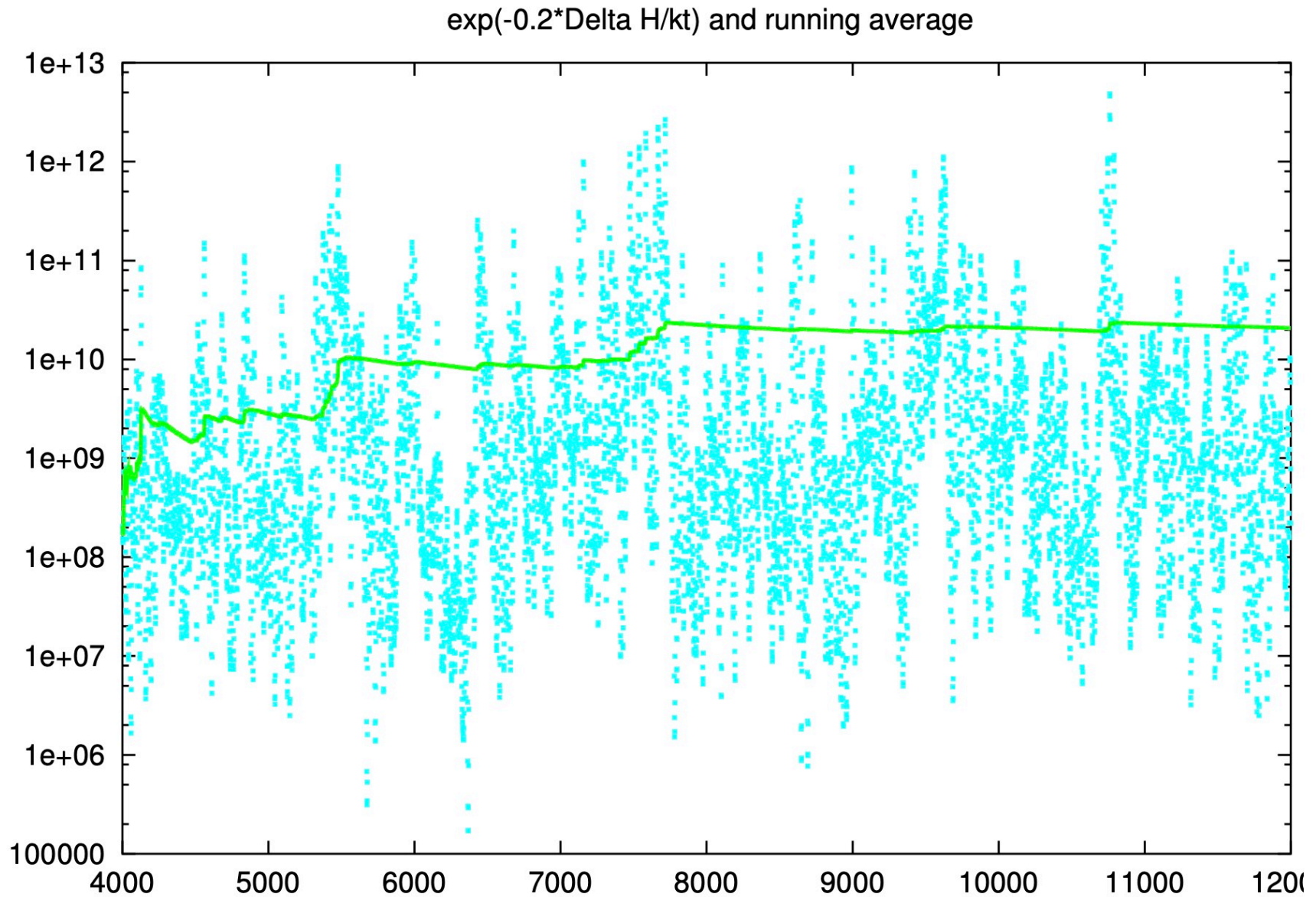
Running Averages

$\exp(-\Delta H/kt)$ and running average



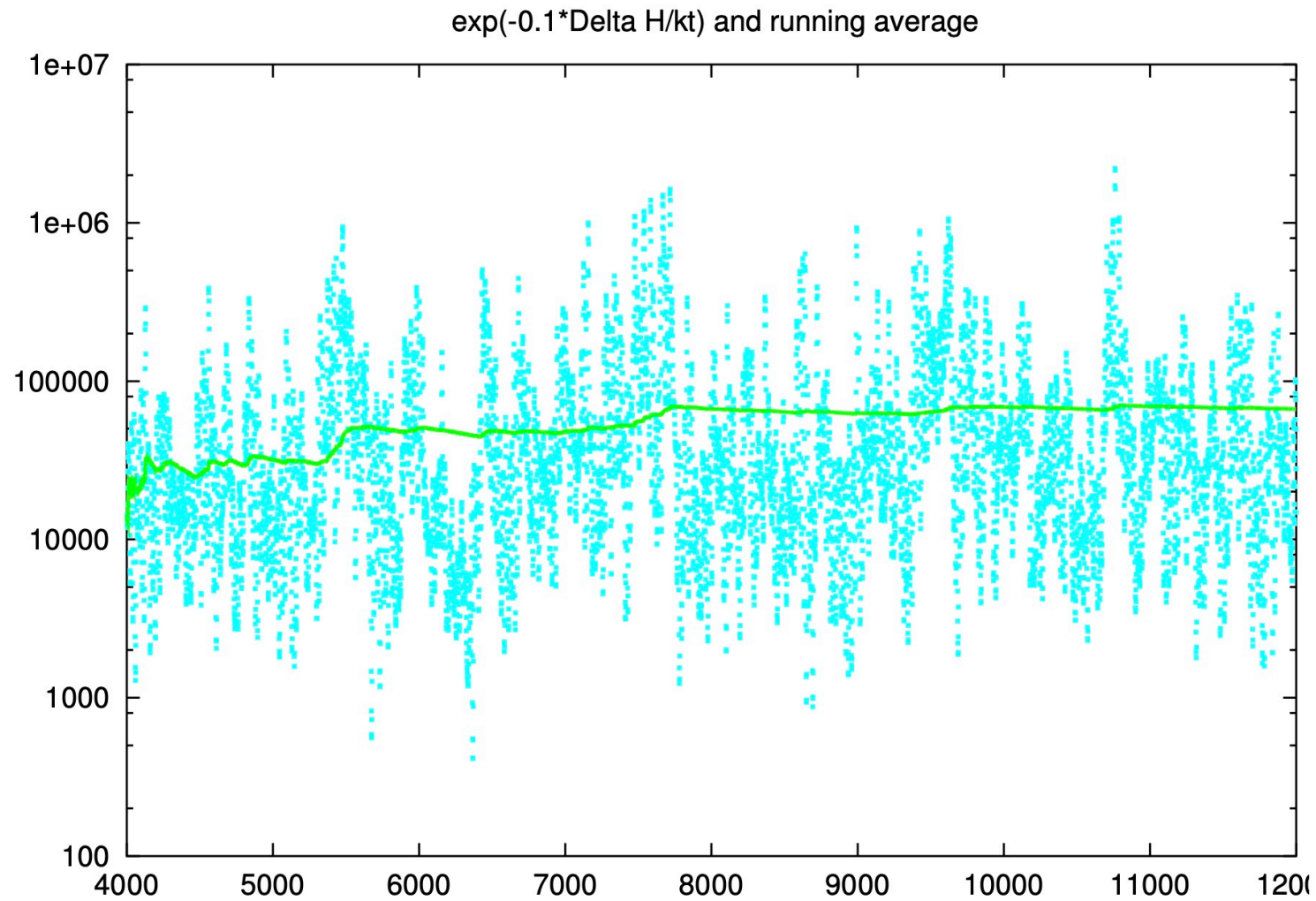
few configurations dominate the average!

Running Averages



still a few configurations dominate the average!

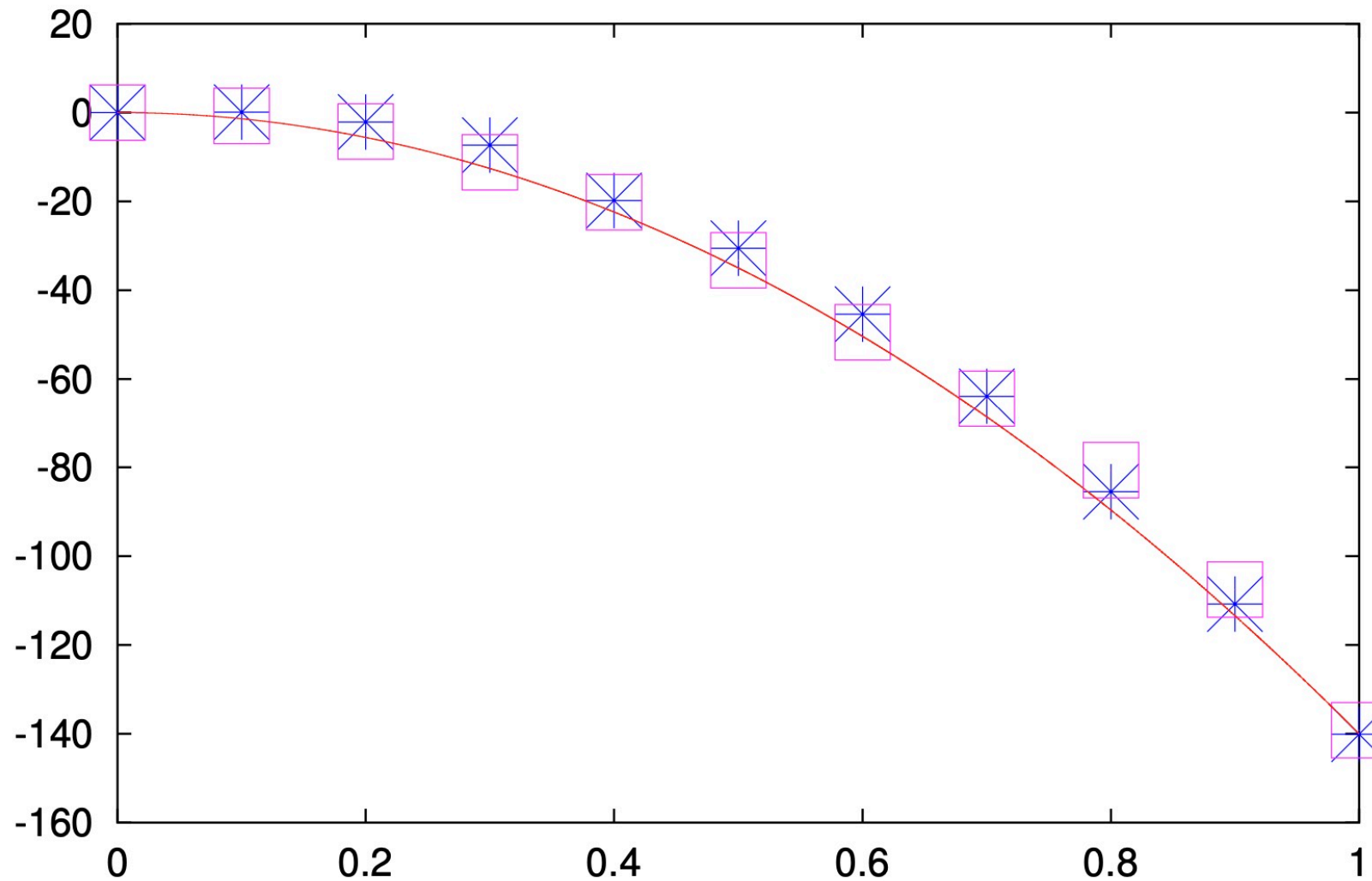
Running Averages



many configurations determine average

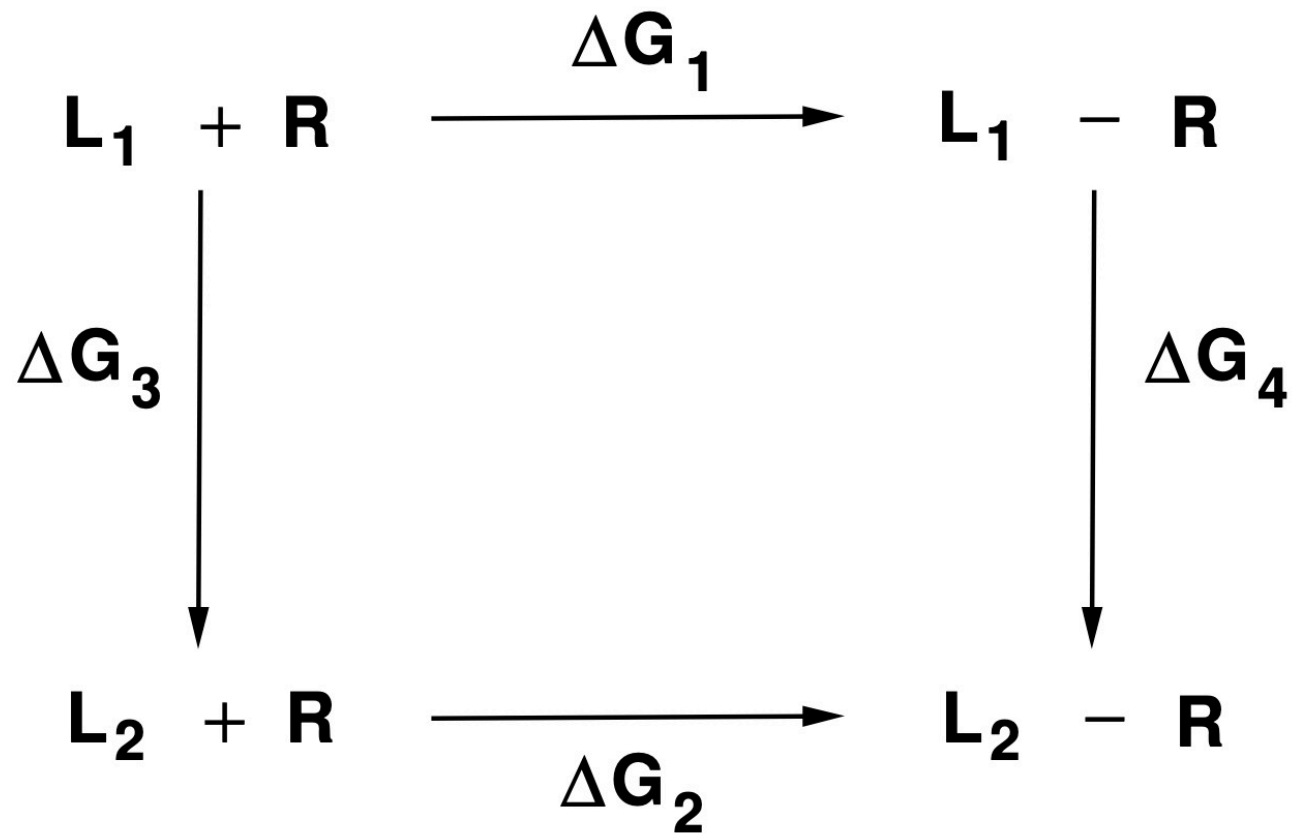
Running Averages

forward and backward free energy differences for each step (in kT)



- realisation of the Born model

Mutation Simulations



$$\Delta G_2 - \Delta G_1 = \Delta G_4 - \Delta G_3$$

frequently used technique in drug design studies

Atomistic Simulation?

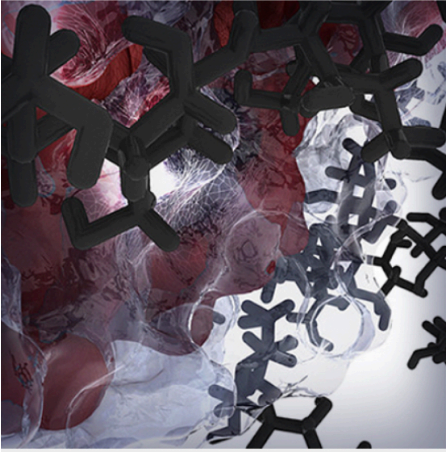
- **modeling** of real physical (chemical / biological) systems
- a subdiscipline of **statistical mechanics**
- calculates **thermodynamic** properties
- follows the 'real' trajectory of a system
- studies structure, evolution thereof

- 6 • **Molecular Dynamics**: studies the **dynamics** of
 - individual atoms
 - **molecules**
 - clusters
 - structural elements
 -
 - parts of living cells
 - **materials**
- chemical **reaction dynamics**

Protein Agglomeration ('Crowding')

VIRTUAL ISSUE

Protein Crowding and Stability



Protein structure determines function, but proteins are shaped *in vivo* where biology takes place in crowded environments: a significant molecular diversity of macromolecules. This leads to a distracting mechanism against which biological function must be maintained. Crowding effects and may also facilitate aggregation. How biological environments affect protein stability.

[Read more](#)

- mechanisms of protein aggregation
- RDFs
- specific interactions
- classical MD

Protein Agglomeration

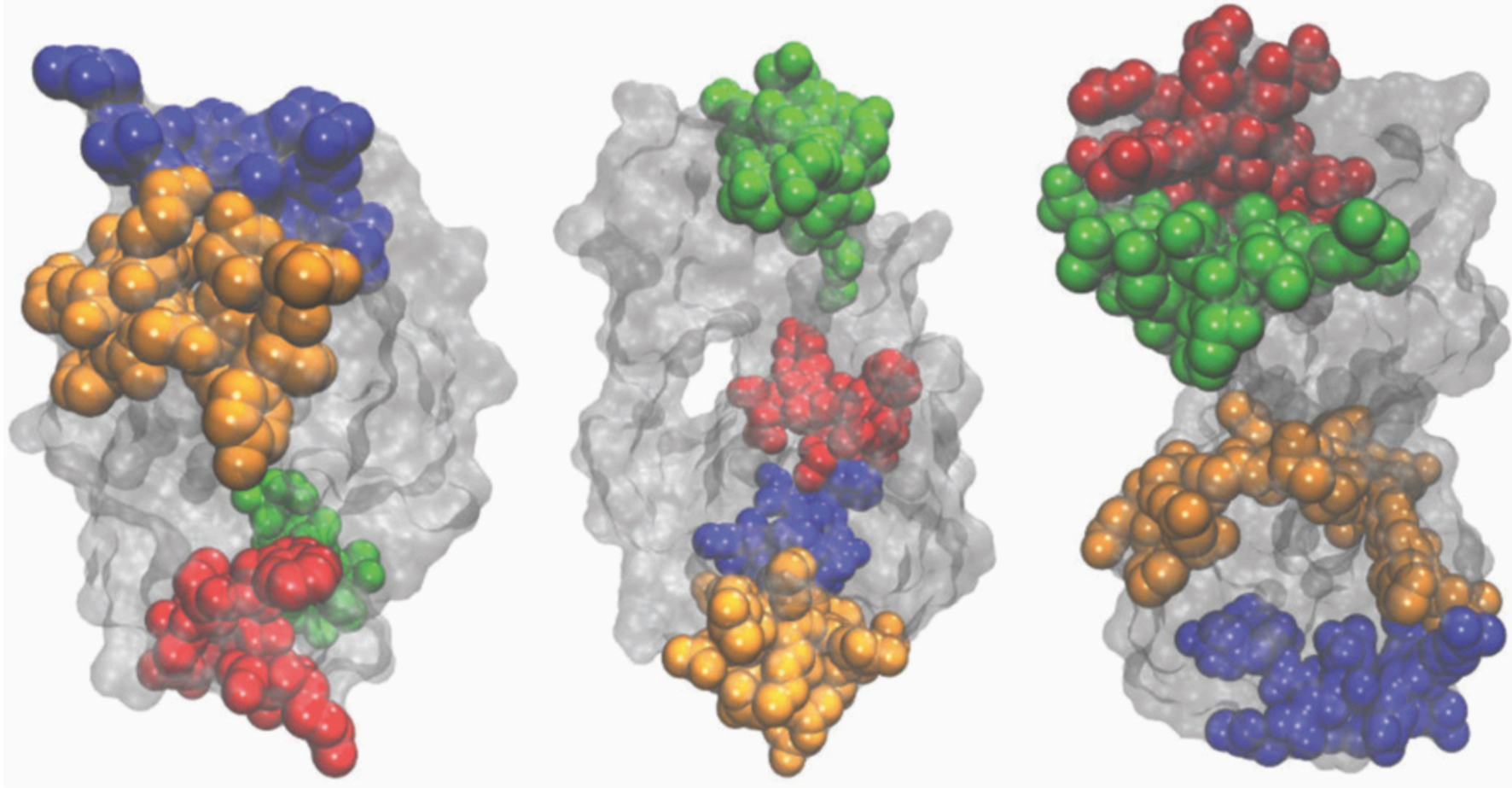
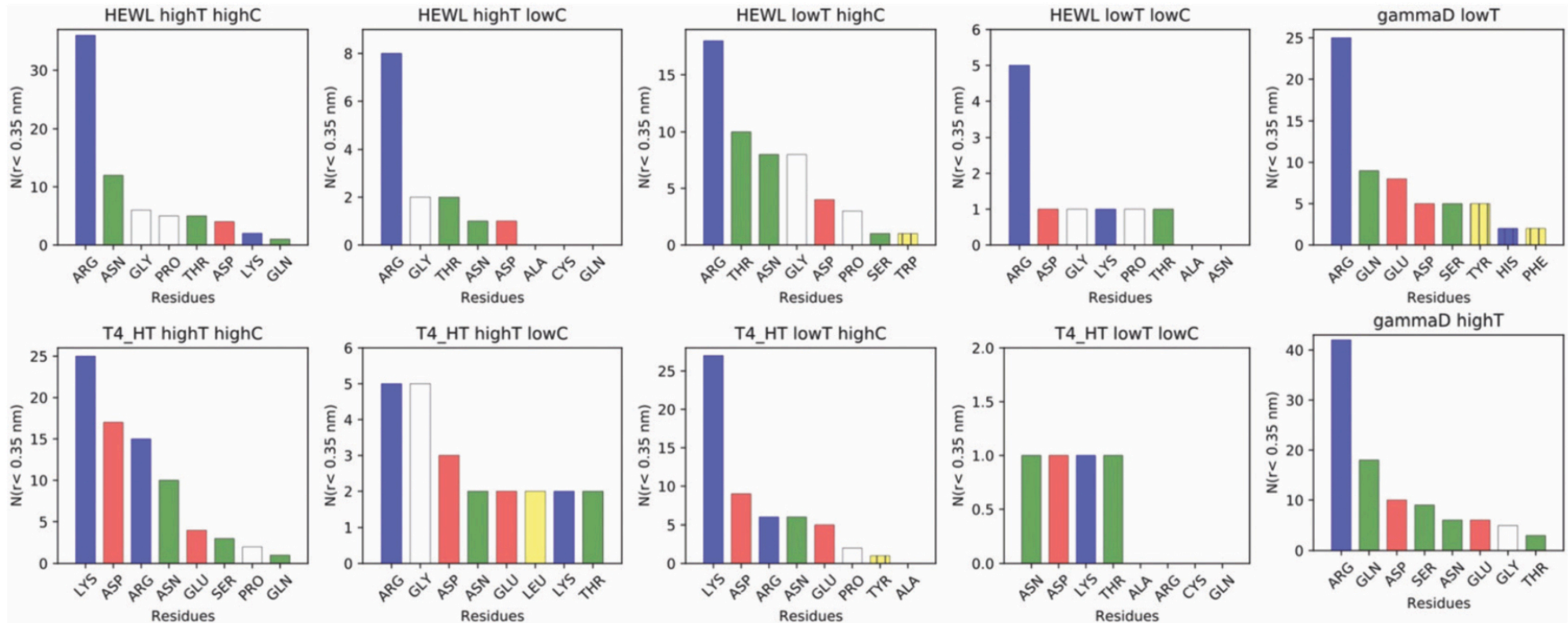


Fig. 5 Four regions of protein molecules that most frequently participate in protein–protein contacts for HEWL (left), T4 WT* (middle), and γ -D crystallin (right).

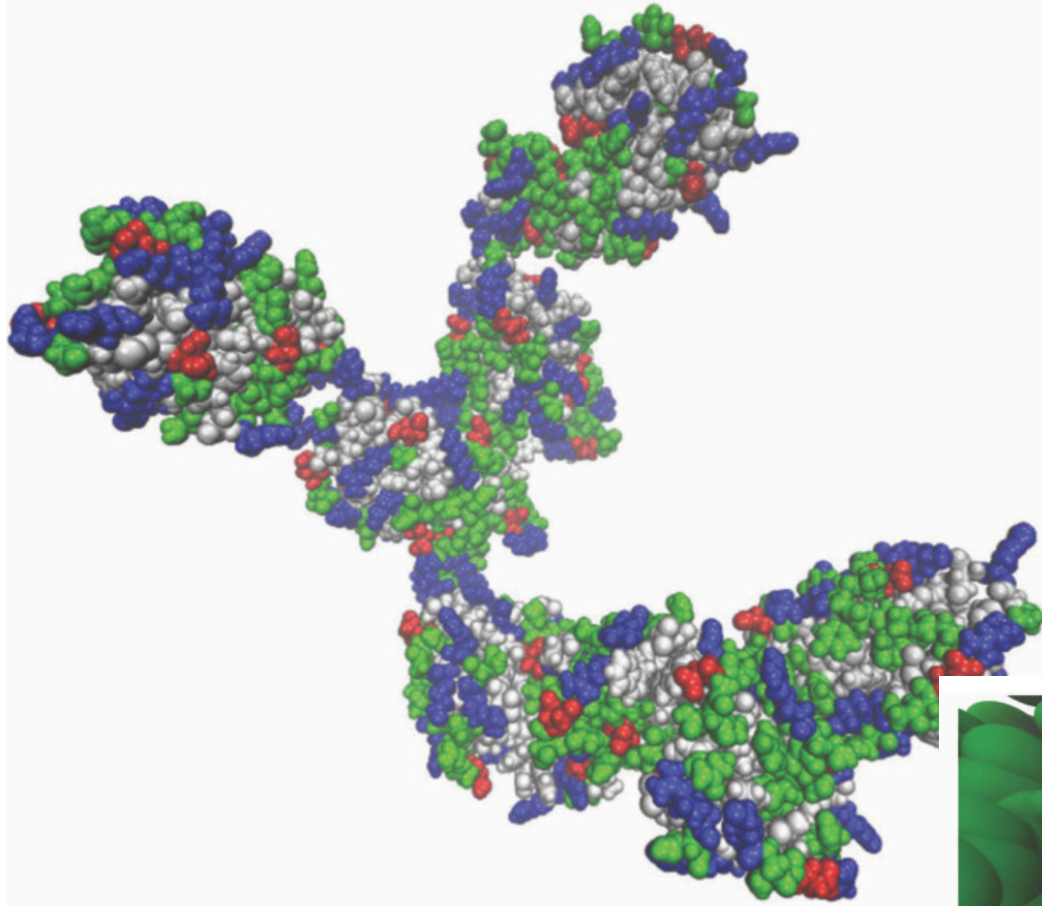
S. Brudar, J. Gujt, E. Spohr, and B. Hribar-Lee, PCCP 23, 415, (2021)

Protein Agglomeration

statistics on cluster-‘initializing’ polar amino acid groups

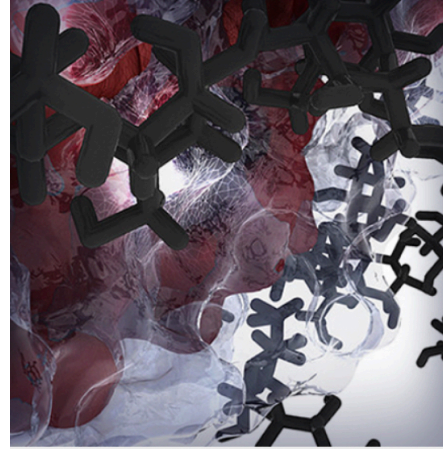


Protein Agglomeration



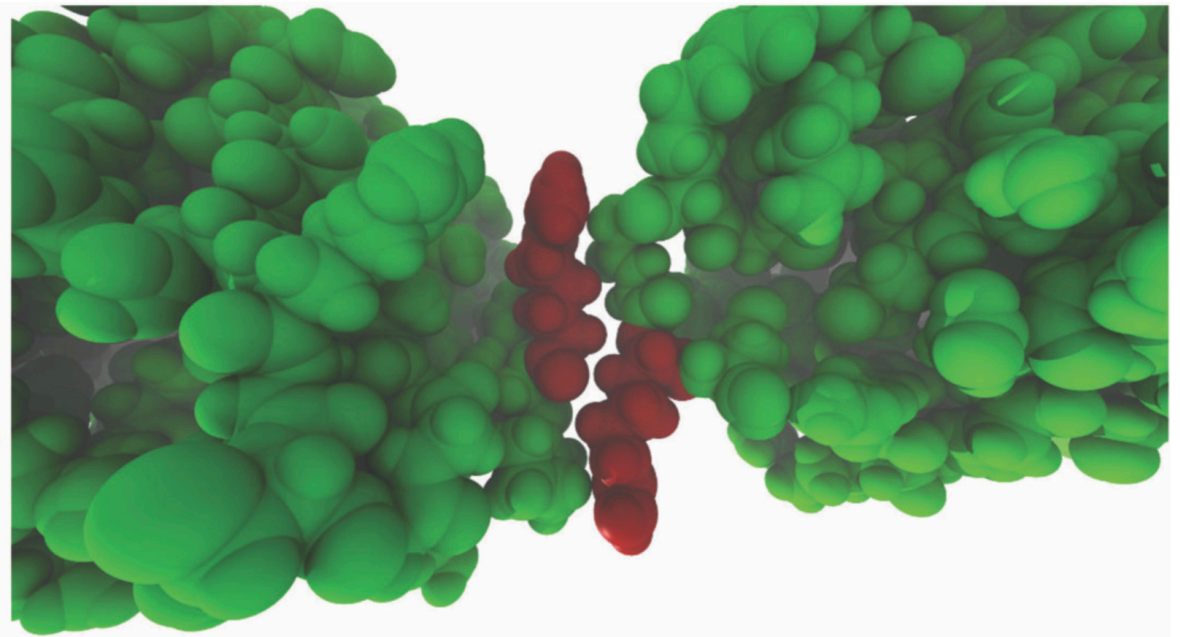
VIRTUAL ISSUE

Protein Crowding and Stability



Protein structure determines function, but proteins are shaped *in vivo* where biology takes place in crowded environments: a significant molecular diversity of macromolecules. This leads to a distracting molecular environment against which biological function must be maintained. Crowding effects and may also facilitate aggregation. How biological environments affect protein stability.

[Read more](#)



Arginine Solution

OPLS force field

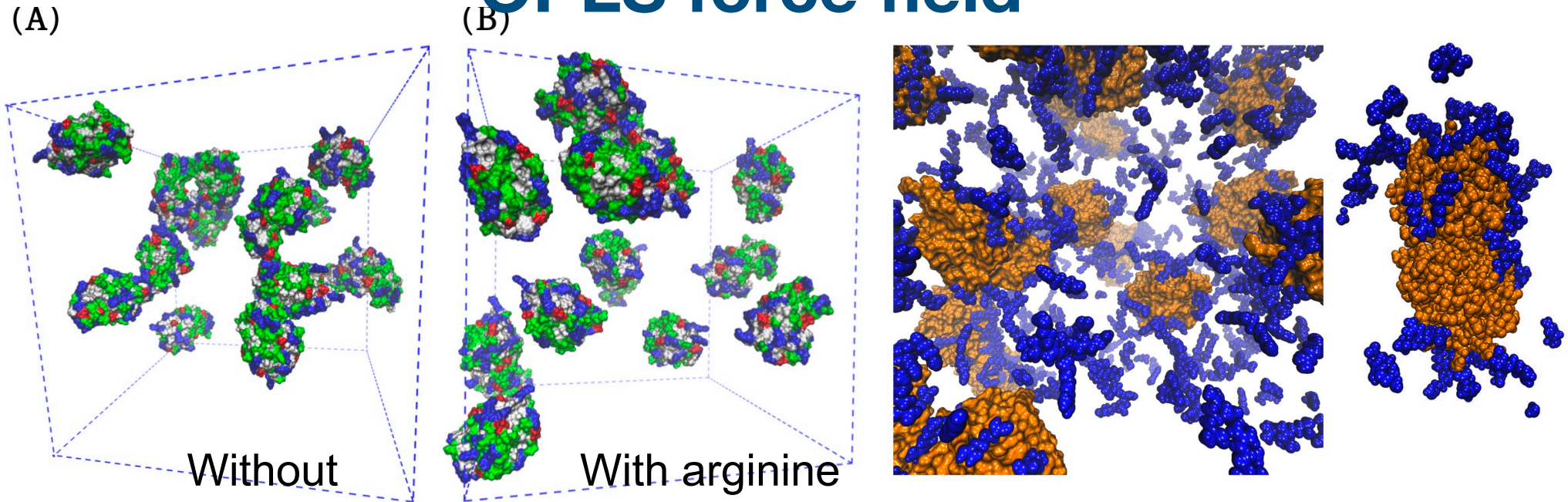


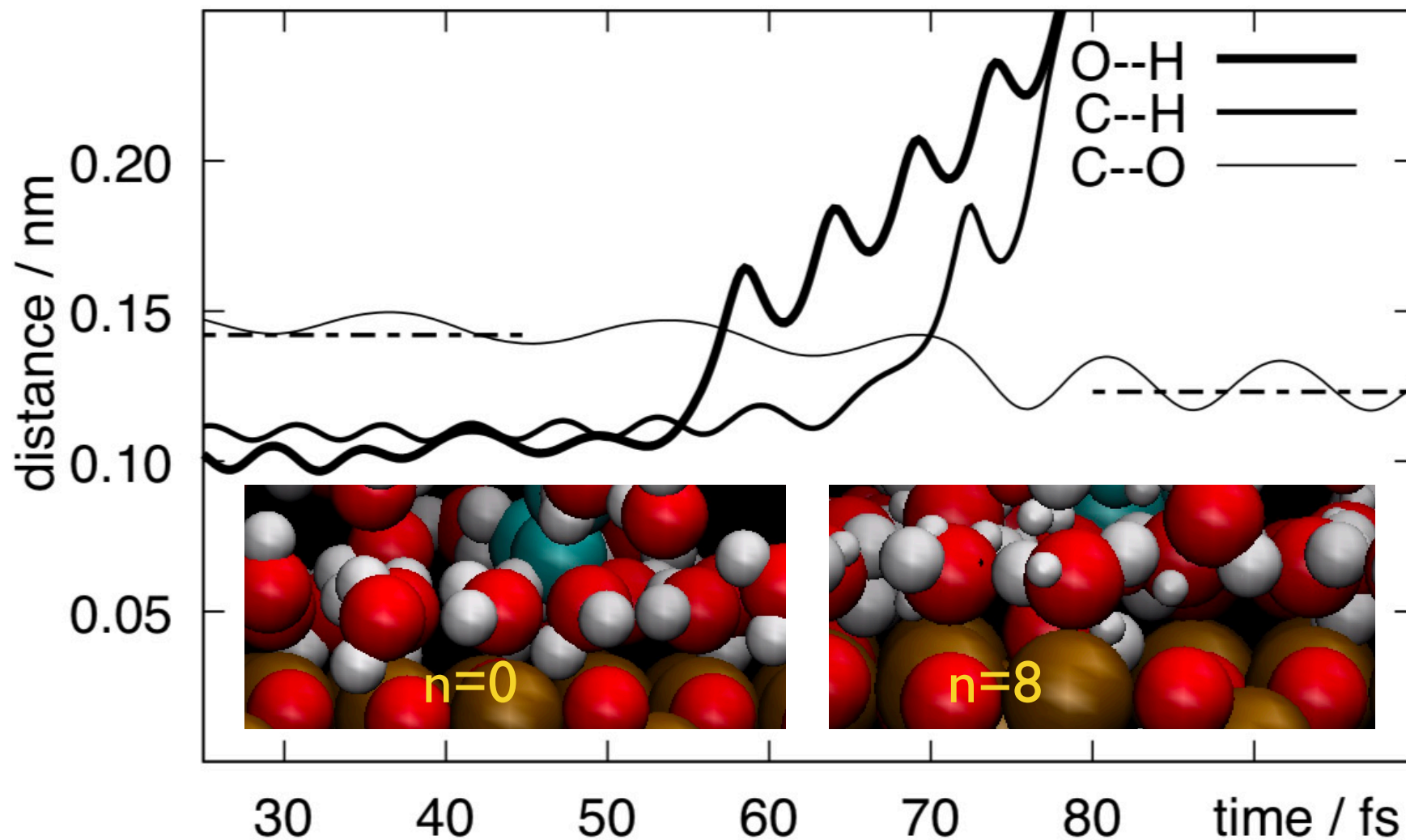
Figure 4. Snapshots from the computer simulations showing just HEWL molecules (A) without the presence of free arginine and (B) with added 0.3 M arginine.

- Experiments show that the addition of arginine increases the stability of the HEWL solutions.
- The computer simulation results indicate that arginine molecules tend to self-associate
- If arginine residues are located on the protein surface, the free arginine molecules stay in their vicinity and prevent protein molecules from “connecting” through them to form clusters.

Atomistic Simulation?

- **modeling** of real physical (chemical / biological) systems
- a subdiscipline of **statistical mechanics**
- calculates **thermodynamic** properties
- follows the ‘real’ trajectory of a system
- studies structure, evolution thereof
- **Molecular Dynamics**: studies the dynamics
 - individual atoms
 - **molecules**
 - clusters
 - structural elements
 -
 - parts of living cells
 - **materials**
- **7** chemical **reaction dynamics**

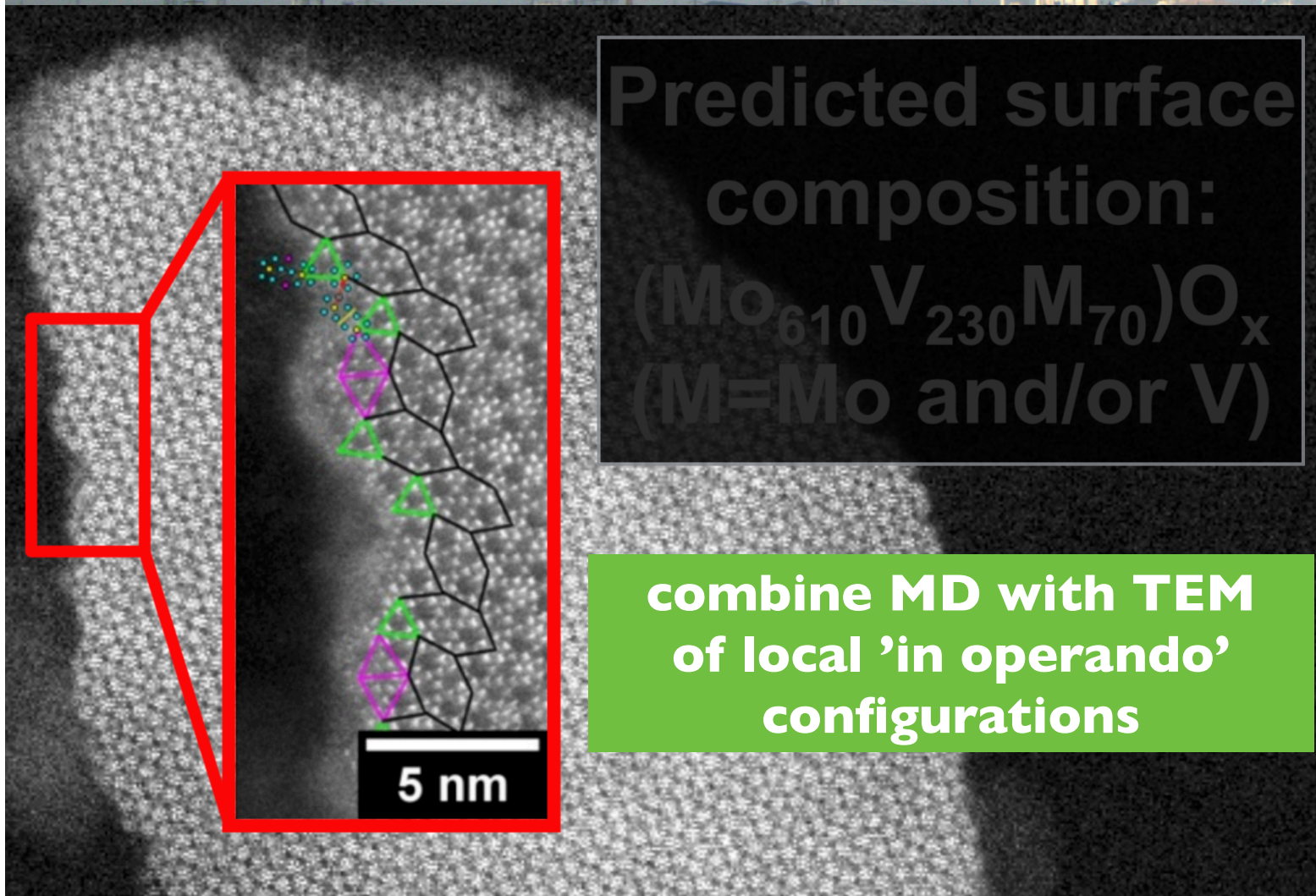
'Electrochemical' Approach



Computer Experiment

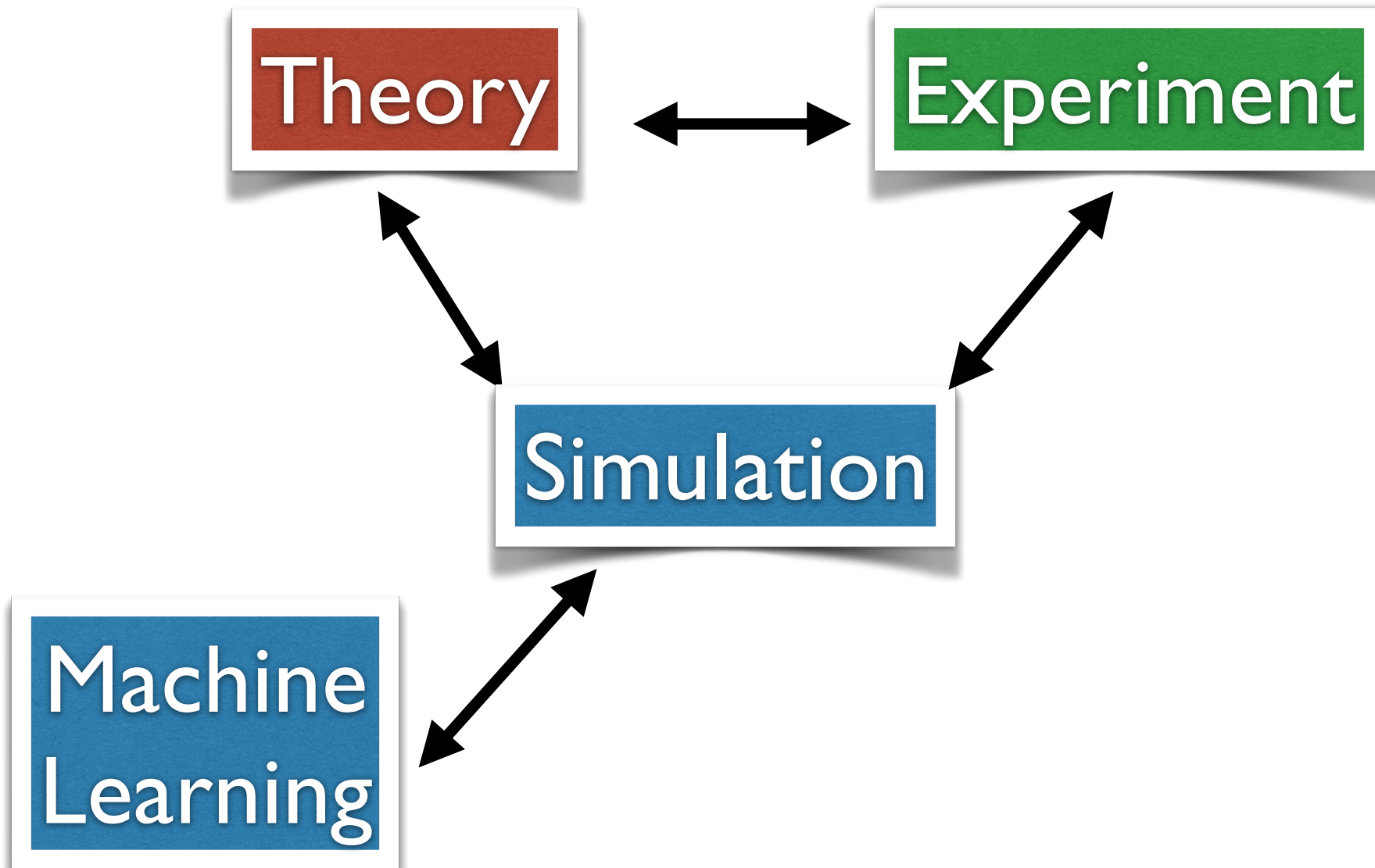
- remove n H atoms (from water molecules or from surface OH)

Perspective



- aqueous environment and proton transfer modulate (electro,photo)catalytic rxns
- local structure is essential
- transition metal atoms: 'buffer' for electrons
- aqueous phase: 'buffer' for protons

Perspective





Thank You!