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2018, June 5 & 8

#### Thermodynamic properties

time averages of thermodynamic quantites

 correspond to ensemble averages (ergodic theorem)

 some quantities – evaluated directly

$$U = \langle E \rangle_t$$

fluctuations – may determine interesting properties: isochoric heat capacity:

$$C_{V} = \left(\frac{\partial U}{\partial T}\right)_{V} = \frac{\sigma_{E}^{2}}{k_{\rm B}T^{2}} = \frac{\langle E^{2} \rangle - \langle E \rangle^{2}}{k_{\rm B}T^{2}}$$

- elegant way from a single simulation to heat capacity

#### General note on averaging

simulated 2 MD trajectories  $\rightarrow$  two sets of 1000 values of A each perform averaging of A separately  $\rightarrow \mu_1 \pm \sigma_1$  and  $\mu_2 \pm \sigma_2$ 

how to average over the whole ensemble over 2000 values?

• 
$$\mu = \frac{1}{2}(\mu_1 + \mu_2)$$

- what about the std. deviation σ?
- hint: make use of  $\sigma^2 = \left< A^2 \right> \left< A \right>^2$
- solution: for each set, perform averaging of A as well as  $A^2$ , then it is safe to average the averages  $\langle A^2 \rangle = \frac{1}{2} \left( \langle A^2 \rangle_1 + \langle A^2 \rangle_2 \right)$ which leads to  $\sigma$

# Single molecule in solvent

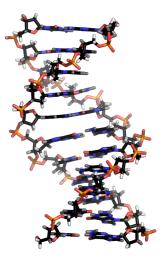
concentrating on the dissolved molecule – protein, DNA,...

#### average structure

- arithmetic mean of coordinates from snapshots along MD trajectory

$$\vec{r_i} = \frac{1}{N} \sum_{n=0}^{N} \vec{r_i}^{(n)}$$

- clear, simple, often reasonable



#### Average structure

#### Possible problems:

- rotation of the entire molecule no big issue
  - RMSD fitting of every snapshot to the starting structure what is RMSD? see on the next slide...
- $\blacksquare$  freely rotatable single bonds  $CH_3$ 
  - all 3 hydrogens collapse to a single point
  - no problem ignore hydrogens
- molecule does not oscillate around a single structure
  - several available minima of free energy
  - possibly averaging over multiple sections of trajectory

#### Dynamic information

#### root mean square deviation (RMSD)

of structure in time tfrom a suitable reference structure  $\vec{r}^{\text{ref}}$ 

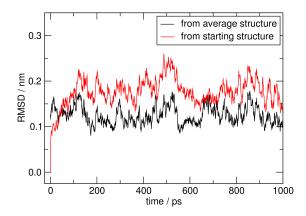
$$\mathsf{RMSD}(t) = \sqrt{rac{1}{N}\sum_{i=1}^{N} \left| ec{r_i}(t) - ec{r_i}^{\mathsf{ref}} 
ight|^2}$$

- follows the development of structure in time
- reference structure starting or average geometry
- also possible comparison with another geometry of interest DNA: A- and B-like; proteins: α-helix and extended β

RMSD fitting – finding such a translation + rotation that minimizes the RMSD from the reference structure

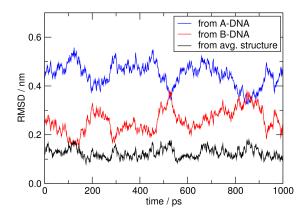
#### Root mean square deviation

# RMSD of non-hydrogen atoms of a DNA oligonucleotide from given geometries



#### Root mean square deviation

# RMSD of non-hydrogen atoms of a DNA oligonucleotide from given geometries



#### Root mean square deviation

**B-DNA** 



average structure





### Magnitude of structural fluctuation

root mean square fluctuation (RMSF) of position of every single atom averaged along MD trajectory

$$\mathsf{RMSF}_i = \sqrt{\left\langle \left| ec{r_i} - \left\langle ec{r_i} 
ight
angle 
ight|^2 
ight
angle}$$

- may be converted to B-factor

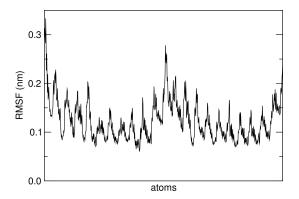
$$B_i = \frac{8}{3}\pi^2 \cdot \mathrm{RMSF}_i^2$$

observable in diffraction experiments (X-ray...)
 contained in structure files deposited in the PDB
 comparison of simulation with X-ray may be difficult

Analysis of the simulation

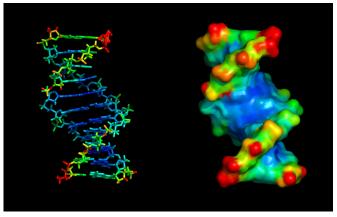
#### Root mean square fluctuation

RMSF of atomic positions in DNA oligonucleotide



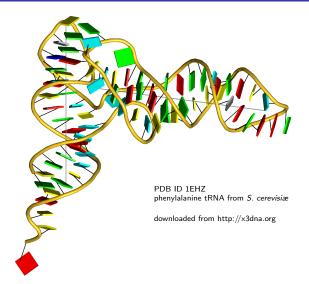
#### Root mean square fluctuation

RMSF of atomic positions in DNA oligonucleotide



(blue < green < yellow < red)

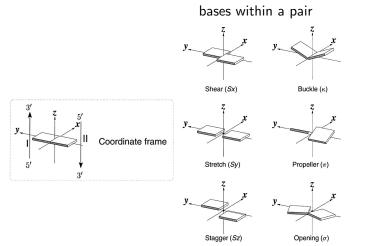
### Structure of double-helical nucleic acids



└─ Structural data

#### Structure of double-helical nucleic acids

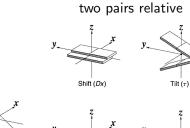
Helical parameters

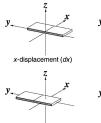


Structural data

#### Structure of double-helical nucleic acids

#### Helical parameters pair in a helix





y-displacement (dy)









Slide (Dy)

Roll  $(\rho)$ 



v

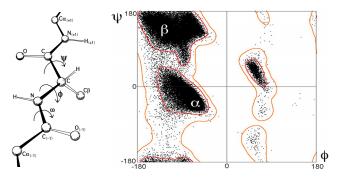




### Structure of peptides and proteins

#### Ramachandran plot

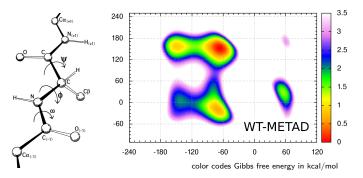
- 2D histogram of dihedrals  $\phi$  and  $\psi$  along the backbone
- different regions correspond to various second. structures
- may be generated easily in simulation software packages



### Structure of peptides and proteins

#### Ramachandran plot

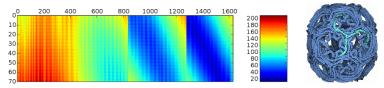
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## Structure of peptides and proteins

#### Distance matrix

- distances of amino-acid residues,
  - represented e.g. by centers of mass or by  $\mathsf{C}^\alpha$  atoms
- either time-dependent or averaged over trajectory
- bioinformatics



distance matrix between two chains (horiz. and vertical axes) shows contacts between secondary structure elements

PDB ID 1XI4, clathrin cage lattice, April 2007 Molecule of the Month

 $http://www2.warwick.ac.uk/fac/sci/moac/people/students/peter\_cock/python/protein\_contact\_map$ 

### Structure of fluids

- example pure argon or water different situation
  - many molecules, which are all equally important

#### radial distribution functions

- describe how the molecular density varies
   as a function of the distance from one particular molecule
- spherical shell of thickness  $\delta r$  at a distance  $r: \ \delta V \approx 4\pi r^2 \cdot \delta r$
- count the number of molecules in this shell: n
- divide by  $\delta V$  to obtain a 'local density' at distance r

### Structure of fluids

example - pure argon or water - different situation

- many molecules, which are all equally important

#### radial distribution functions

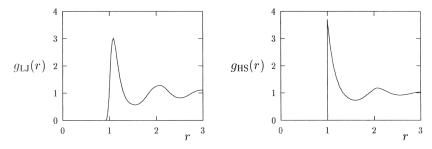
pair distribution function

$$g(r) = \frac{n/\delta V}{\rho} = \frac{n}{4\pi r^2 \cdot \delta r} \cdot \frac{1}{\rho}$$

- probability to find a molecule in distance r from ref. mol. - division by the macroscopic density - normalization

## Pair distribution function

#### Lennard-Jones fluid near the triple point and hard-sphere fluid



reprinted from Nezbeda, Kolafa and Kotrla 1998

### Pair distribution function

- g(r) vanishes on short distances molecules cannot intersect
- high peak van der Walls radius, closest-contact distance (even though hard spheres do not have any attraction!)
   much more likely to find this distance in LJ or HS than in IG
- longer distances a few shallow minima and maxima, converges to unity – uniform probability as in IG

Fourier transform of g(r) – structure factor S

- quantifies the scattering of incoming radiation in the material
- measured in diffraction experiments (X-ray, neutron)

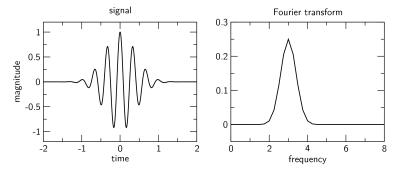
$$S(\vec{q}) = \frac{1}{N} \left\langle \sum_{j} \sum_{k} \exp\left[-i \cdot \vec{q} \cdot (\vec{r_j} - \vec{r_k})\right] \right\rangle$$

#### intermission: Fourier transform

FT describes which frequencies are present in a function (of time) – decomposes f(t) into a 'sum' of periodic oscillatory functions

$$F(\omega) = \int_{-\infty}^{\infty} f(t) \cdot \exp\left[-\mathrm{i}\,\omega t\right] \,\mathrm{d}t$$

note that 
$$\exp\left[-i\omega t\right] = \cos\left[\omega t\right] - i\sin\left[\omega t\right]$$

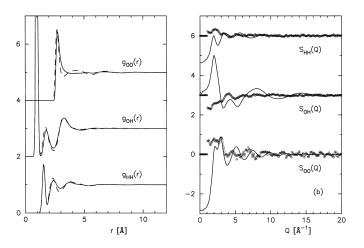


Structural data

#### Pair distribution function

g(r) and S(q) of water

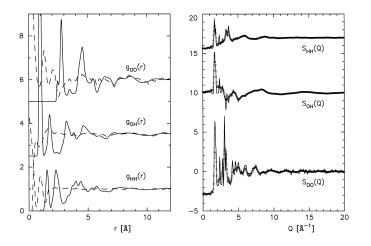
(Soper, Chemical Physics 2000)



#### Pair distribution function

g(r) and S(q) of ice lh at 220 K and 1 bar

(Soper, Chemical Physics 2000)



### Pair distribution function

Importance – not only information about the structure calculation of thermodynamic properties possible using potential energy u(r) and force f(r) of a molecule pair

corrections to the IG values of total energy and pressure (EOS!):

$$E - \frac{3}{2}Nk_{\rm B}T = 2\pi N\rho \int_0^\infty r^2 \cdot u(r) \cdot g(r) \, \mathrm{d}r$$
$$P - \rho \, k_{\rm B}T = -\frac{2\pi}{3}\rho^2 \int_0^\infty r^3 \cdot f(r) \cdot g(r) \, \mathrm{d}r$$

(as long as pairwise additivity of forces can be assumed)

### Equilibration

- 'preliminary' simulation to reach the termodynamic equilibrium
- goal stable thermodynamics properties (no drift)
- usually E<sub>pot</sub>, T, p, in NPT also ρ
   evaluated by program readily and written to output
- structure has to be taken care of, too
- start often artificially regular (crystal-like) structure, which should be washed out during equilibration

Monitoring the equilibration

#### Structural parameters

translational order – Verlet's order parameter

$$\lambda = \frac{\lambda_x + \lambda_y + \lambda_z}{3}, \qquad \lambda_x = \frac{1}{N} \sum_{i=1}^N \cos\left[\frac{4\pi x_i}{a}\right] \quad \text{etc.}$$

a – edge of the unit cell ideal crystal:  $\lambda = 1$ disordered structure:  $\lambda$  fluctuates around 0

mean squared displacement from initial position

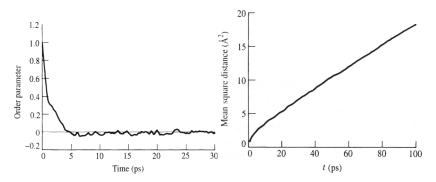
$$\mathsf{MSD} = \frac{1}{N} \sum_{i=1}^{N} |\vec{r_i}(t) - \vec{r_i}(0)|^2$$

should increase gradually in fluid with no specific structure would oscillate about a mean value in a solid

Monitoring the equilibration

#### Structural parameters

#### equilibration of liquid argon followed by $\lambda$ and MSD



Reprinted from Leach: Molecular Modelling

### Correlation functions

two physical quantities x and y may exhibit correlation

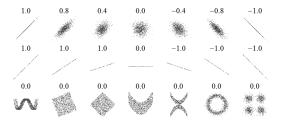
- indicates a relation of *x* and *y*, opposed to independence
- quantification several kinds of correlation functions
- Pearson correlation coefficients
  - describe linear relationship between x and y
  - quantities fluctuate around mean values  $\langle x 
    angle$  and  $\langle y 
    angle$
  - consider only the fluctuating part, i.e.  $x \langle x \rangle$  and  $y \langle y \rangle$
  - introduce correlation coefficient  $\rho_{xy}$

$$\rho_{xy} = \frac{\langle (x - \langle x \rangle) \cdot (y - \langle y \rangle) \rangle}{\sqrt{\langle (x - \langle x \rangle)^2 \rangle \cdot \langle (y - \langle y \rangle)^2 \rangle}} = \frac{\operatorname{cov}(x, y)}{\sigma_x \cdot \sigma_y}$$

cov(x, y): covariance of x and y

#### Correlation functions

#### (not necessarily linear) correlation of two quantities and the coresponding correlation coefficients



Downloaded from Wikipedia

### Correlation functions

MD – values of a quantity x as a function of time

- at some point in time, the value of x may be correlated with the value of x at an earlier time point
- described by autocorrelation function (ACF)

$$c_{x}(t) = \frac{\langle x(t) \cdot x(0) \rangle}{\langle x(0) \cdot x(0) \rangle} = \frac{\int x(t') x(t'+t) dt'}{\int x^{2}(t') dt'}$$

 correlation of the same property x at two time points separated by t, normalized to takes values between -1 and 1

### Autocorrelation of velocity

autocorrelation function – quantifies 'memory' of the system, or how quickly the system 'forgets' its previous state

#### velocity autocorrelation function

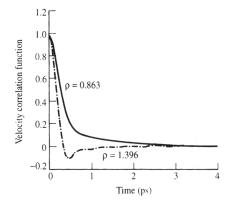
tells how closely the velocities of atoms at time t resemble those at time 0
usually averaged over all atoms i in the simulation

$$c_{
m v}(t) = rac{1}{N}\sum_{i=1}^{N}rac{\langleec v_i(t)\cdotec v_i(0)
angle}{\langleec v_i(0)\cdotec v_i(0)
angle}$$

• typical ACF starts at 1 in t = 0 and decreases afterwards

### Autocorrelation of velocity

ACF of velocity in simulations of liquid argon (densities in  $g \cdot cm^{-3}$ )



Reprinted from Leach: Molecular Modelling

lower  $\rho$  – gradual decay to 0

higher  $\rho$  – ACF comes faster to 0

- even becomes negative briefly
- 'cage' structure of the liquid
- one of the most interesting achievements of early simulations

### Autocorrelation of velocity

time needed to lose the autocorrelation whatsoever

- correlation time or relaxation time:

$$\tau_{\mathbf{v}} = \int_0^\infty c_{\mathbf{v}}(t) \, \mathrm{d}t$$

may help to resolve certain statistical issues: when averaging over time the properties of system, it is necessary to take uncorrelated values if the property is dynamical (related to v), we can take values of the property separated by  $\tau_v$ 

### Autocorrelation of velocity

connection between velocity ACF and transport properties

Green–Kubo relation for self-diffusion coefficient *D*:

$$D = rac{1}{3} \int_0^\infty \langle ec{v}_i(t) \cdot ec{v}_i(0) 
angle_i \, \mathrm{d}t$$

- interesting observable quantities
- important to be able to calculate them from MD
- there is yet another way from simulation to D
   Einstein relation for D using the MSD

$$D = \frac{1}{6} \lim_{t \to \infty} \frac{\left\langle |\vec{r_i}(t) - \vec{r_i}(0)|^2 \right\rangle_i}{t}$$

NB: Fick's laws of diffusion  $J = -D\frac{\partial\phi}{\partial x}$ ,  $\frac{\partial\phi}{\partial t} = D\frac{\partial^2\phi}{\partial x^2}$ 

## Autocorrelation of dipole moment

velocity – property of a single atom other quantities – need to be evaluated for whole system

total dipole moment:

$$ec{\mu}_{ ext{tot}}(t) = \sum_{i=1}^{N} ec{\mu}_i(t)$$

Λ/

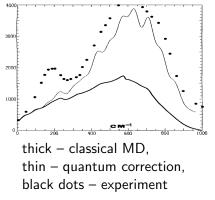
ACF of total dipole moment:

$$c_{\mu}(t) = rac{\langle ec{\mu}_{ ext{tot}}(t) \cdot ec{\mu}_{ ext{tot}}(0) 
angle}{\langle ec{\mu}_{ ext{tot}}(0) \cdot ec{\mu}_{ ext{tot}}(0) 
angle}$$

- related to the vibrational spectrum of the sample
- IR spectrum may be obtained as Fourier transform of  $c_{\mu}(t)$

## Autocorrelation of dipole moment

#### IR spectra for liquid water from simulations



B. Guillot, J. Phys. Chem. 1991

no sharp peaks at well-defined frequencies (as in gas phase)

rather – continuous bands – liquid absorbs frequencies in a broad interval

frequencies – equivalent to the rate of change of total dipole moment

#### Principal component analysis

covariance analysis on the atomic coordinates along MD trajectory = principal component analysis (PCA), or essential dynamics

3*N*-dim. covariance matrix *C* of atomic coordinates  $r_i \in \{x_i, y_i, z_i\}$ 

$$C_{ij} = \langle (r_i - \langle r_i \rangle) \cdot (r_j - \langle r_j \rangle) \rangle_t \quad \text{or} \\ C_{ij} = \langle \sqrt{m_i} (r_i - \langle r_i \rangle) \cdot \sqrt{m_j} (r_j - \langle r_j \rangle) \rangle_t \quad (\text{mass-weighted})$$

diagonalization  $\rightarrow$ 

eigenvalues – may be expressed as quasi-harmonic frequencies eigenvectors – principal or essential modes of motion

- analogy of normal modes of vibration

- first few - largest eigenvalues, lowest frequencies

- global, collective motions, many atoms involved

## intermission: diagonalization of a matrix

is a process of finding

eigenvectors *a* with corresponding eigenvalues  $\alpha$  of a square  $(n \times n)$  matrix A:

 $\mathcal{A} \cdot \mathbf{a} = \alpha \cdot \mathbf{a}$ 

(here: eigenvectors are column vectors)

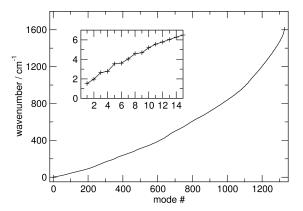
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why the name?
```

if eigenvectors are arranged into a matrix  $\mathbf{a} = (a_1 a_2 \dots a_n)$ then  $\mathbf{a}^{-1} \cdot \mathcal{A} \cdot \mathbf{a}$  is a diagonal matrix

- $\blacksquare$  symmetric or Hermitian matrix  $\rightarrow$  all eigenvalues are real
- computational cost of diagonalization:  $\mathcal{O}(n^3)$

### Principal component analysis

Double-stranded DNA oligonucleotide - lowest frequencies

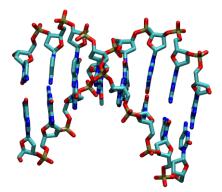


10 ns simulation of a double-helical DNA 11-nucleotide 691 atoms, of which 445 non-hydrogen ightarrow 1329 vib. modes

└─ Time-dependent properties

#### Principal component analysis

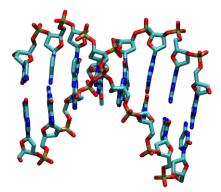
DNA octamer, eigenvector 1



└─ Time-dependent properties

#### Principal component analysis

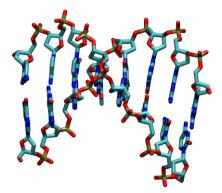
DNA octamer, eigenvector 2



└─ Time-dependent properties

#### Principal component analysis

DNA octamer, eigenvector 3



#### Principal component analysis

DNA – the modes are the same as expected for a flexible rod

- 2 bending modes around axes perpendicular to the principal axis of the DNA, and a twisting mode
- PCA gives an idea of what the modes of motion look like – additionally – basis for thermodynamic calculations – vibrational frequencies may lead to configurational entropy

#### Principal component analysis

Binding domain of a glutamate receptor protein

