#### Biomolecular modeling II

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## Molecular dynamics simulation

how to get things moving

# Equations of motion

$$m \cdot \ddot{\vec{r}} = \vec{F}$$

#### ordinary differential equations of second order

- have to be solved numerically
- solution proceeds in discreet steps of length  $\Delta t$
- numerical integration starts at time t<sub>0</sub>, where the initial conditions are specified - the positions r<sub>0</sub> and the velocities v<sub>0</sub>
- calculations of forces at  $\vec{r_0}$  to get accelerations  $\ddot{\vec{r}_0}$
- ullet then, an integrator calculates ec r and ec v at time  $t_0+\Delta t$
- accelerations  $\rightarrow$  step  $\rightarrow$  accelerations  $\rightarrow$  step  $\rightarrow \dots$

#### Verlet integration method

for the development of the method:

take a virtual step in positive time and in 'negative' time, and apply Taylor expansion up to second order:

$$\begin{aligned} r(t + \Delta t) &= r(t) + \dot{r}(t) \cdot \Delta t + \frac{1}{2} \ddot{r}(t) \cdot \Delta t^2 \\ r(t - \Delta t) &= r(t) - \dot{r}(t) \cdot \Delta t + \frac{1}{2} \ddot{r}(t) \cdot \Delta t^2 \end{aligned}$$

add both equations – eliminate the velocity  $\dot{r}$ :

$$r(t + \Delta t) = 2 \cdot r(t) - r(t - \Delta t) + \ddot{r}(t)\Delta t^{2}$$
$$\ddot{r}(t) = a(t) = \frac{F(t)}{m} = -\frac{1}{m}\frac{\partial V}{\partial r}(t)$$

## Verlet integration method

another, equivalent formulation - velocity Verlet

$$\begin{aligned} r(t + \Delta t) &= r(t) + v(t) \cdot \Delta t + \frac{1}{2}a(t) \cdot \Delta t^2 \\ v(t + \Delta t) &= v(t) + \frac{1}{2}(a(t) + a(t + \Delta t)) \cdot \Delta t \end{aligned}$$

yet another – Leap-frog

$$\begin{aligned} \mathsf{v}(t + \frac{1}{2}\Delta t) &= \mathsf{v}(t - \frac{1}{2}\Delta t) + \mathsf{a}(t) \cdot \Delta t \\ \mathsf{r}(t + \Delta t) &= \mathsf{r}(t) + \mathsf{v}(t + \frac{1}{2}\Delta t) \cdot \Delta t \end{aligned}$$

both: better numerical precision than Verlet normal form

#### $\Delta t$ – crucial parameter

Let us say: we want to obtain a trajectory over a time interval T

- we perform M steps
- we have to evaluate the forces on atoms  $M = T/\Delta t$  times

Computational cost of the calculation of forces

- major computational effort
- determines how many steps we can afford to make

#### $\Delta t$ – crucial parameter

- we neglect contributions in  $\Delta t^3$  and higher orders  $\rightarrow$  error per step in the order of  $\Delta t^3$
- keep the step short  $\rightarrow$  make the error small but need too many steps to simulate certain time T
- make the step long  $\rightarrow$  cut computational cost but increase the error and decrease stability
- compromise needed

#### $\Delta t$ – crucial parameter

- fastest motion hydrogen atoms, period around 10 fs
- rule of thumb stable integration with  $\Delta t \leq \frac{1}{10}$  fastest period
- practically, Δt of 1 fs is used, increase to 2 fs possible with a special treatment of bonds
- 1M calculations of forces needed for a trajectory of 1 ns
- large systems multi-ns simulations routinely,  $\mu$ s possible

Total energy with a force field

$$V(R^N) =$$

$$= \frac{1}{2} \sum_{i} k_{i} (r_{i} - r_{i}^{0})^{2} + \frac{1}{2} \sum_{j} k_{j}^{\vartheta} (\vartheta_{j} - \vartheta_{j}^{0})^{2} + \frac{1}{2} \sum_{n} V_{n} \cdot \cos[n\omega - \gamma_{n}] \\ + \sum_{i}^{N} \sum_{j=i+1}^{N} \left\{ 4\varepsilon_{ij} \left( \left( \frac{\sigma_{ij}}{r_{ij}} \right)^{12} - \left( \frac{\sigma_{ij}}{r_{ij}} \right)^{6} \right) + \frac{1}{4\pi\varepsilon_{0}} \frac{q_{i}q_{j}}{r_{ij}} \right\}$$

and get forces as derivatives with respect to atomic coordinates:

$$F_i^x = -\frac{\partial V}{\partial x_i}$$

### Verlet integration method

$$r(t + \Delta t) = 2 \cdot r(t) - r(t - \Delta t) + \ddot{r}(t)\Delta t^{2}$$
$$\ddot{r}(t) = a(t) = \frac{F(t)}{m} = -\frac{1}{m}\frac{\partial V}{\partial r}(t)$$

and choose an appropriate time step  $\Delta t$ 

### Temperature and pressure

#### what you simulate is what you would measure

### **Energies and temperature**

Solution of equations of motion - conserves total / internal energy



### Energies and temperature

Solution of equations of motion – conserves total / internal energy

what we need – to control basic simulation parameters – temperature and possibly pressure

significance of temperature

- determines which structures of the system are accessible
- different dynamics at high and at low temperatures

## Energies and temperature

high E – multiple different structural 'classes' are reached low E – restricted available structures



difference  $E - E_{pot}$  corresponds to  $E_{kin}$  and temperature

# Isolated system

- exchanges with surroundings neither energy (heat / work) nor matter (particles)
- total energy of system:  $E = E_{kin} + E_{pot} = const$
- individually,  $E_{kin}$  and  $E_{pot}$  fluctuate in the course of time as they are being transformed into each other
- is what we get when using the Verlet method for a molecule

# Isolated system

kinetic theory of gases  $\rightarrow$  relation of  $E_{\rm kin}$  and temperature:

$$\langle E_{
m kin} 
angle = rac{3}{2} N k T$$
  
where  $\langle E_{
m kin} 
angle = rac{1}{2} \sum_{i} m_{i} \langle v_{i}^{2} 
angle$ 

'local' T – fluctuates in time; may differ between parts of system

Isolated and closed system

experimental setup (a test tube with a sample)

- usually in thermodynamic equilibrium with the surroundings
- temperature of system = temperature of suroundings



# Closed system

- thermal contact of system with surroundings
- exchange of energy in the form of heat until the temperature of surroundings is reached
- canonical ensemble
- velocity / speed of atoms Maxwell-Boltzmann distribution

### Canonical ensemble



#### Idea for constant temperature – scale the velocities

'instantaneous temp.' T in a simulation deviates from target  $T_{ref}$ :

$$T(t) = rac{2}{3} rac{E_{kin}(t)}{Nk_{B}} 
eq T_{ref}$$
  $E_{kin}(t) = rac{1}{2} \sum_{i} m_{i} \cdot v_{i}(t)^{2}$ 

how to control T? scale the velocities by a certain factor  $\lambda$ , so that temperature is modified and  $T_{ref}$  is reached:

$$T_{\text{ref}} = \frac{1}{\frac{3}{2}Nk_{\text{B}}} \cdot \frac{1}{2} \sum_{i} m_{i} \left( \lambda \cdot \mathbf{v}_{i} \right)^{2} = \lambda^{2} \cdot \frac{1}{\frac{3}{2}Nk_{\text{B}}} \cdot \frac{1}{2} \sum_{i} m_{i} v_{i}^{2} = \lambda^{2} \cdot T$$

thus: scale all velocities by  $\lambda=\sqrt{T_{\rm ref}/T}~\rightarrow~T_{\rm ref}$  reached exactly

Problem: rescaling of velocities affects the 'natural' dynamics of the molecular system too strongly  $\rightarrow$  unstable simulation

### Berendsen thermostat

How to avoid the drastic changes to the dynamics? adjust velocities more smoothly, in the direction of  $T_{\rm ref}$ 

• temperature changes between two time steps according to

$$\Delta T = rac{\Delta t}{ au} \left( T_{\mathsf{ref}} - T 
ight)$$

- rate of change of T (due to the change of velocities) is proportional to the deviation of actual T from  $T_{ref}$
- constant of proportionality relaxation time  $\tau$

#### Berendsen thermostat

• velocities are scaled by  $\lambda$ :

$$T_{\text{new}} = T + \Delta T = T + \frac{\Delta t}{\tau} (T_{\text{ref}} - T)$$
$$\lambda = \sqrt{\frac{T_{\text{new}}}{T}} = \sqrt{1 + \frac{\Delta t}{\tau} \left(\frac{T_{\text{ref}}}{T} - 1\right)}$$

- usually: au = 0.1 10 ps
- T will fluctuate around the desired value  $T_{ref}$

#### Berendsen thermostat

Question - is the statistical thermodynamics correct?

- in other words is distribution of velocities correct (M–B)?
- it is important to make sure that canonical ensemble is sampled because every thermodynamic quantity A is obtained as an ensemble average:

$$\langle A \rangle = \frac{1}{Z} \int \rho(\vec{r}) \cdot A(\vec{r}) \, \mathrm{d}\vec{r}$$

- $\bullet\,$  if sampling is wrong  $\rightarrow\,$  wrong density  $\rho\,\rightarrow\,$  wrong averages
- answer: it is not correct may even lead to problems like ....

#### Berendsen thermostat

## The Flying Ice Cube: Velocity Rescaling in Molecular Dynamics Leads to Violation of Energy Equipartition

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ABSTRACT: This article describes an unexpected phenomenon encountered during MD simulations: velocity rescaling using standard protocols can systematically change the proportion of total kinetic energy (KE) found in motions associated with the various degrees of freedom. Under these conditions, the simulation violates the principle of equipartition of energy, which requires a mean kinetic energy of RT/2 in each degree of freedom. A particularly pathological form of this problem occurs if one does not periodically remove the net translation of (and rotation about) the center of mass. In this case, almost all of the kinetic energy is converted into these two kinds of motion, producing a system with almost no kinetic energy associated with the internal degrees of freedom. We call this phenomenon "the flying ice cube." We present a mathematical analysis of a simple diatomic system with two degrees of freedom, to document the origin of the problem. We then present examples from three kinds of MD simulations, one being an in vacuo simulation on a diatomic system, one involving a low resolution model of DNA in vacuo, and the third using a traditional all-atom DNA model with full solvation, periodic boundary conditions, and the particle mesh Ewald method for treating long-range electrostatics. Finally, we discuss methods for avoiding the problem.

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# Nosé–Hoover thermostat

- $\bullet$  generates the correct canonical ensemble  $\rightarrow$  good choice
- conceptionally and mathematically > difficult to understand
- heat bath is treated not as an external element rather as an integral part of the system; is assigned an additional degree of freedom s with fictitious mass Q
- eqns of motion for this extended system (3N + 1 DOF):

$$\ddot{r}_{i} = \frac{F_{i}}{m_{i}} - s \cdot \dot{r}_{i}$$
  
$$\dot{s} = \frac{1}{Q} \left(T - T_{\text{ref}}\right)^{2}$$

#### Introducing pressure

chemical reality – constant pressure rather than constant volume goal – implement such conditions in simulations, too

How to calculate pressure?

- first, calculate virial of force

$$\Xi = -\frac{1}{2} \sum_{i < j} \vec{r}_{ij} \cdot \vec{F}_{ij}$$

 $(\vec{r}_{ij} \text{ distance of atoms } i \text{ and } j, \vec{F}_{ij} - \text{force between them})$ 

$$P = \frac{2}{3V} \cdot (E_{\text{kin}} - \Xi) = \frac{2}{3V} \cdot \left(\frac{1}{2}\sum_{i} m_i \cdot |\vec{v}_i|^2 + \frac{1}{2}\sum_{i < j} \vec{r}_{ij} \cdot \vec{F}_{ij}\right)$$

#### Measuring pressure

T and P in an NPT simulation of a DNA oligomer in water ( $T_{\rm ref} = 300$  K,  $P_{\rm ref} = 1.0$  bar)



# Controlling pressure

#### we can calculate the pressure

 so how do we maintain it at a desired value?
 barostat - algorithm that is equivalent of a thermostat, just that it varies volume of the box instead of velocities

alternatives are available:

- Berendsen barostat
  - direct rescaling of box volume
  - system coupled to a 'force / pressure bath' piston
- Parrinello-Rahman barostat
  - extended-ensemble simulation
  - additional DOF for the piston

# System boundary and the solvent

### Biomolecule in solution

typical MD simulations – molecular system in aqueous solution preferably – make the system as small as possible (reduce cost)

straightforward solution – single molecule of solute (protein, DNA) with a smallest possible number of  $H_2O$  molecules typical – several thousand  $H_2O$  molecules in a box  $n \times n \times n$  nm

issue – everything is close to the surface, while we are interested in a molecule in bulk solvent

# Periodic boundary conditions

- elegant way to avoid these problems
- molecular system placed in a regular-shaped box
- the box is virtually replicated in all spatial directions



# Periodic boundary conditions

- elegant way to avoid these problems
- molecular system placed in a regular-shaped box
- the box is virtually replicated in all spatial directions
- positions (and velocities) of all particles are identical in all replicas, so that we can keep only one copy in the memory
- this way, the system is infinite no surface!
- the atoms near the wall of the simulation cell interact with the atoms in the neighboring replica

# Periodic boundary conditions



# PBC – features

- only coordinates of the unit cell are kept in memory
- atom that leaves the box enters it on the other side
- carefull accounting of the interactions of atoms necessary! simplest way – minimum image convention:
  - an atom interacts with the nearest copy of every other
  - interaction with two different images of another atom, or even with another image of itself is avoided

# PBC – box shape

may be simple – cubic or orthorhombic, parallelepiped (specially, rhombohedron), or hexagonal prism





# PBC – box shape

- ... but also more complicated
  - truncated octahedral or rhombic dodecahedral
  - quite complex equations for interactions & eqns of motion



advantage for simulation of spherical objects (globular proteins) – no corners far from the molecule filled with unnecessary  $\rm H_2O$ 

### PBC – box shape

2D objects – phase interfaces, membrane systems – usually treated in a slab geometry



### Water in biomolecular simulations

most simulations – something in aqueous solutions  $H_2O$  – usually (many) thousands of molecules



## Water in biomolecular simulations

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example - simulation of DNA decanucleotide:

- PBC box  $3.9 \times 4.1 \times 5.6$  nm (smallest meaningful)
- ullet 630 atoms in DNA, 8346 atoms in water and 18 Na $^+$
- concentration of DNA: 18 mmol/L very high!
- of all pair interactions: 86 % are water-water, most of the others involve water

## Water models

most interactions involve  $H_2O$ 

 $\rightarrow$  necessary to pay attention to its description model of water must be simple enough (computational cost)

and accurate enough, at the same time

water models - usually rigid

bond lengths and angles do not vary - constraints

molecule with three sites (atoms in this case), or up to six sites

- three atoms and virtual sites corresponding

to a 'center' of electron density or lone electron pairs



### Water models

#### TIP3P (or SPC)

- most frequently used
- 3 atoms with 3 rigid bonds, charge on every atom (-0.834/+0.417)

 $\bullet$  only the O atom has non-zero LJ parameters (optimization) TIP4P

- negative charge placed on virtual site M rather than on the O
- electric field around the molecule described better

TIP5P

- 2 virtual sites L with negative charges near the O lone pairs
- better description of directionality of H-bonding etc. (radial distribution function, temperature of highest density)