

Computación con GPUs: Aplicación a la simulación de membranas biológicas

Enrique Velasco y Teresa Ruiz

Departamento de Física Teórica de la
Materia Condensada

Universidad Autónoma de Madrid

Molecular Dynamics (MD) simulations

One of the most widely used techniques for studying
condensed systems (fluids, solids,...)

Access to equilibrium and dynamical properties

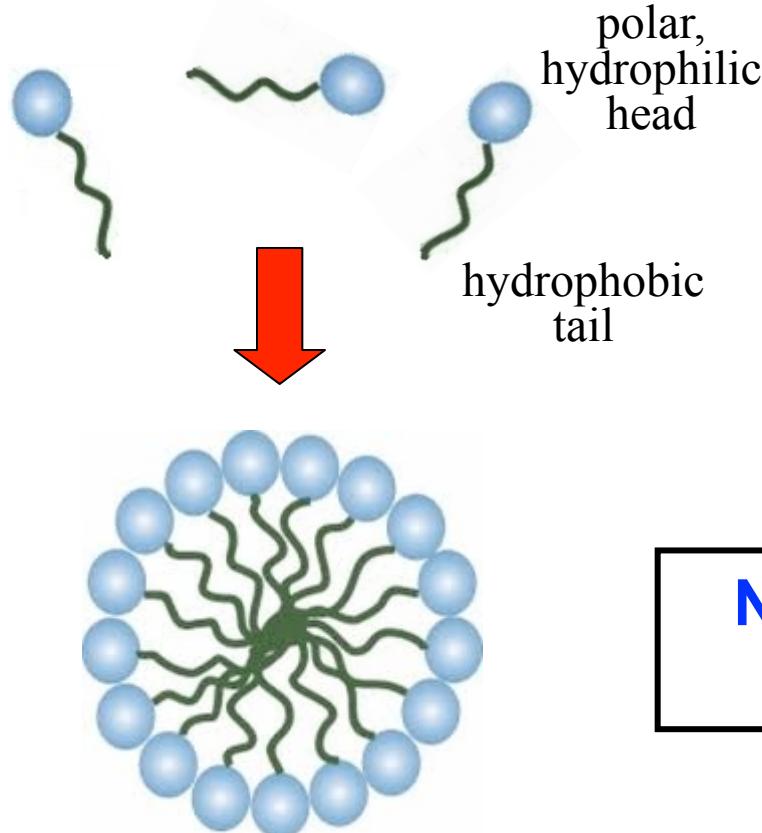
Some old and recent applications:

- Liquids (first applications '40, '60, ...)
- Complex fluids (SOFT MATTER):
liquid crystals, colloids, amphiphilic systems,...
- Biological systems (macromolecules, proteins,...)

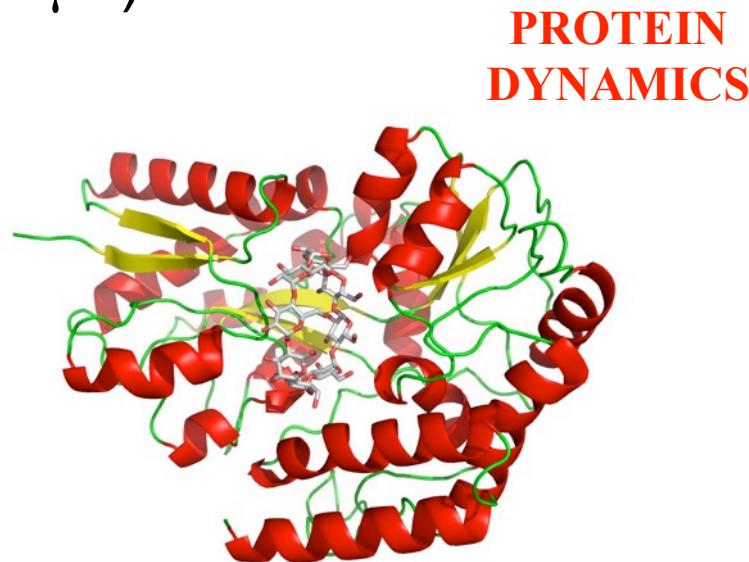


Limiting factors in complex fluid applications: mesoscales

- *Length* scales (not nm but 100 nm – 1 μm)
- *Time* scales (not ps but ns – μs)



MICELLE FORMATION



NEED FOR COARSE
GRAINING

MESOSCOPIC MODELS

EXPERIMENT (real life)

THEORY

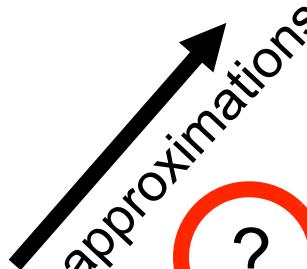
↑
is the
model
good
enough?
|

not always
exact

MODEL



approximations

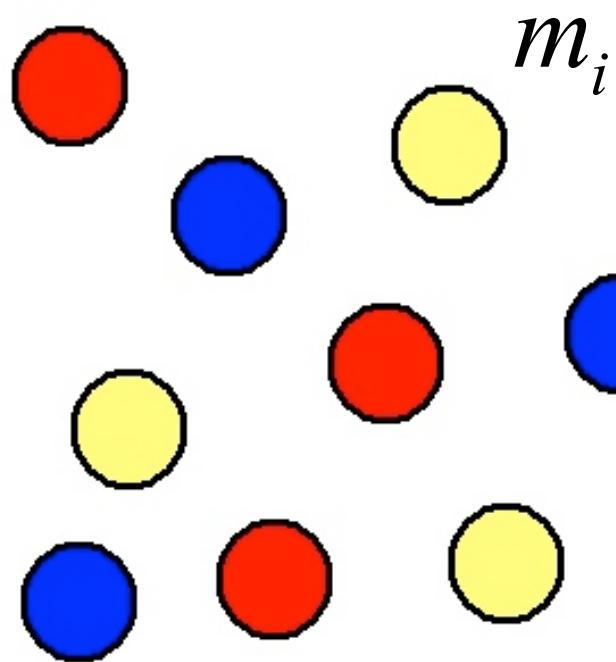
A red circle containing a white question mark, positioned to the right of the MODEL oval.

are theory and
approximations
good enough?



SIMULATION

MD (classical) simulations: basic facts



We solve classical Newton's equations of motion and obtain dynamical trajectories from the forces:

$$\left\{ \begin{array}{l} \frac{d\mathbf{r}_i}{dt} = \mathbf{v}_i, \\ m_i \frac{d\mathbf{v}_i}{dt} = \mathbf{F}_i, \quad \mathbf{F}_i = -\nabla_i \Phi \end{array} \right.$$

Hamiltonian
system (energy
conserved)

$$i = 1, 2, \dots, N$$

- The potential energy Φ has to be specified
- A finite-difference scheme is used

$$\Phi = \Phi(\mathbf{r}_1, \mathbf{r}_2, \dots, \mathbf{r}_N)$$

potential energy
function

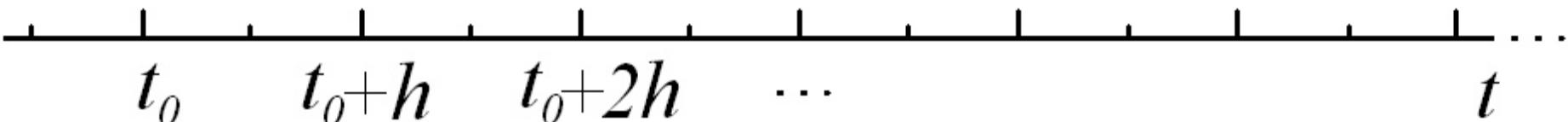
Finite-difference scheme: **Verlet (leap-frog) algorithm**

$$t \circledR t_0, t_1, t_2, \dots \quad t_n = nh, \quad n = 0, 1, 2, \dots$$

h : MD simulation step

$$\left\{ \begin{array}{l} \mathbf{v}_i \left(t_n + \frac{h}{2} \right) = \mathbf{v}_i \left(t_n - \frac{h}{2} \right) + \frac{h}{m_i} \mathbf{F}_i(t_n), \\ \mathbf{r}_i(t_n + h) = \mathbf{r}_i(t_n) + h \mathbf{v}_i \left(t_n + \frac{h}{2} \right) \end{array} \right. \quad i = 1, 2, \dots, N$$

\mathbf{r}_i



\mathbf{v}_i

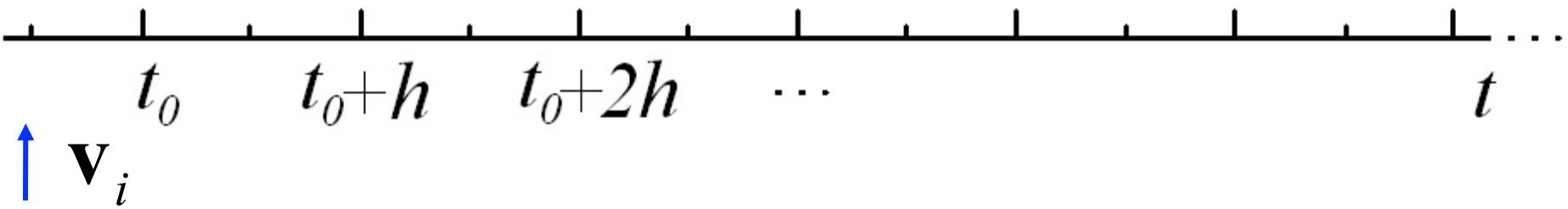
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\mathbf{r}_i ↓

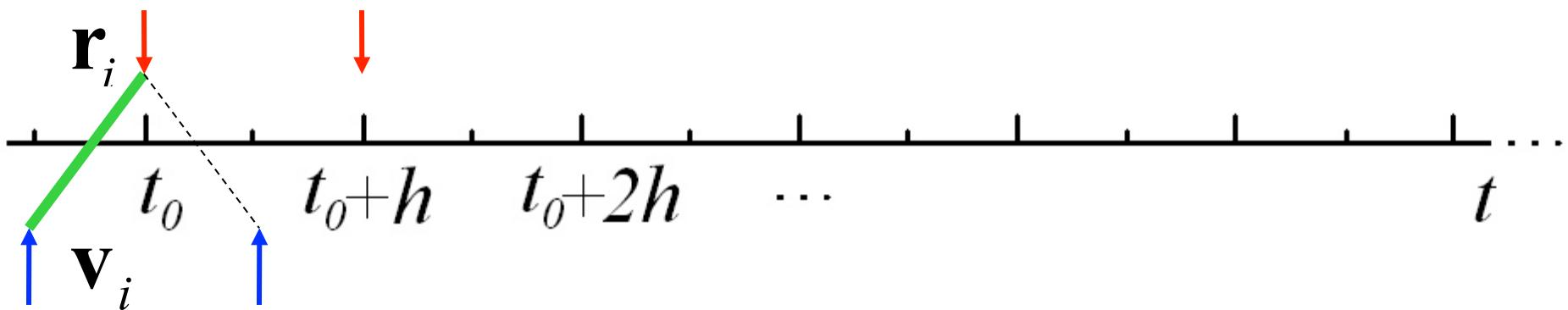


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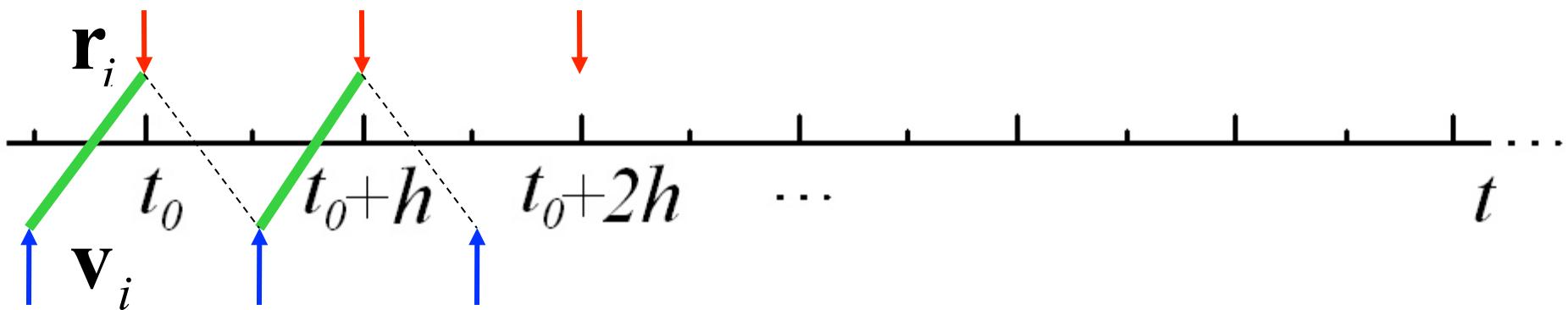


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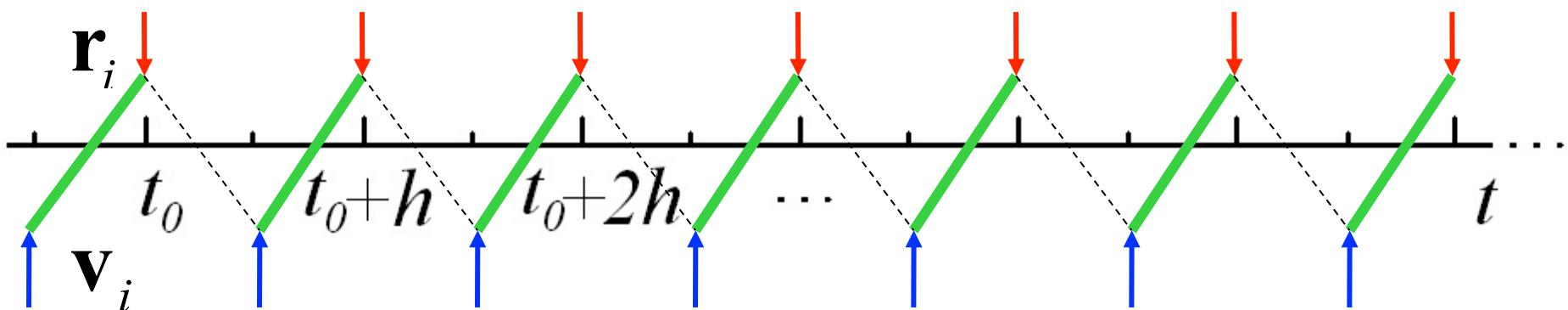


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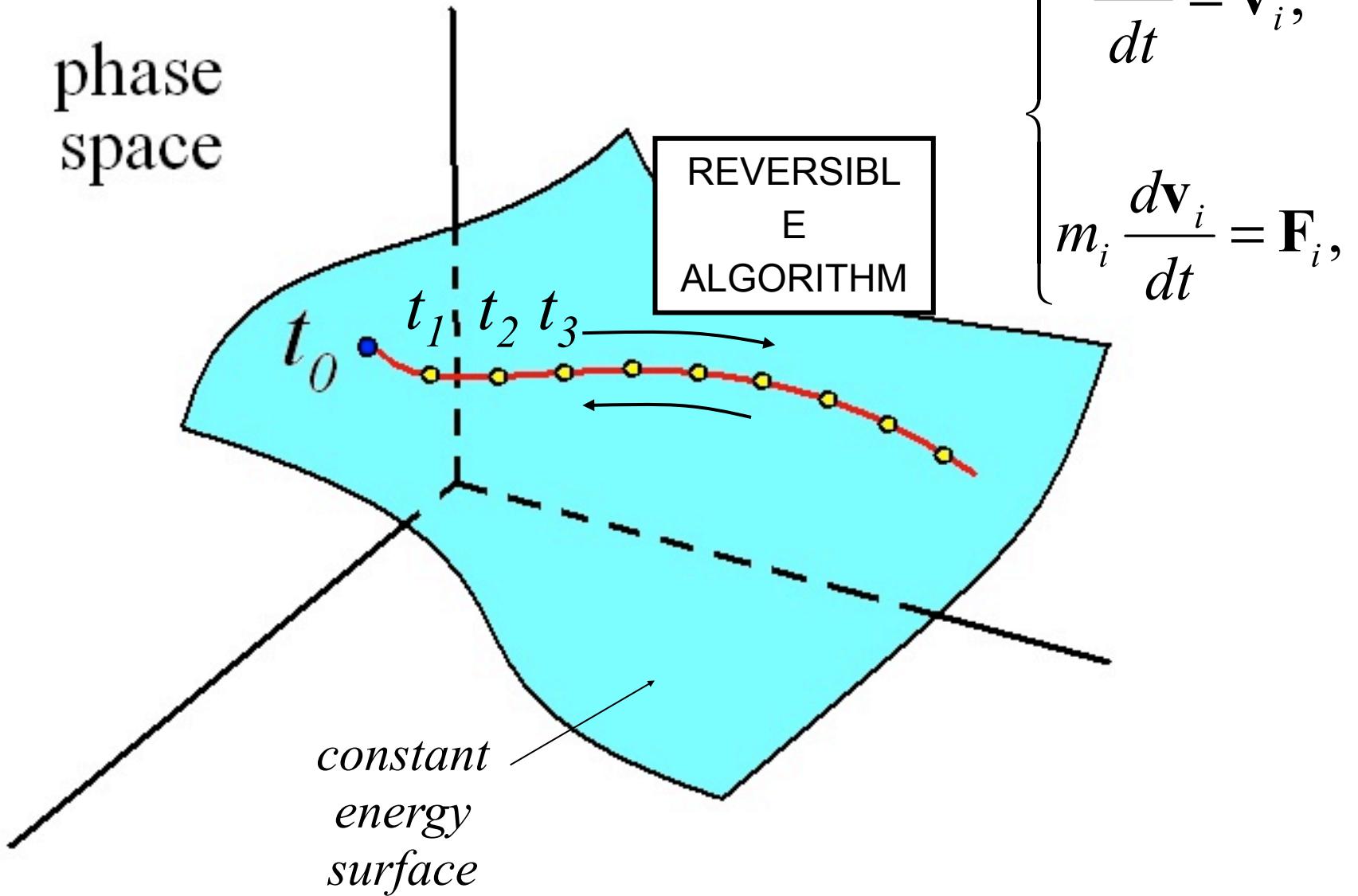
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phase
space



*constant
energy
surface*

$$\left. \begin{array}{l} \frac{d\mathbf{r}_i}{dt} = \mathbf{v}_i, \\ m_i \frac{d\mathbf{v}_i}{dt} = \mathbf{F}_i, \end{array} \right\}$$

$$\begin{pmatrix} \mathbf{v}_i\left(t_n + \frac{h}{2}\right) \\ \vdots \\ \mathbf{r}_i(t_n + h) \end{pmatrix} = J \begin{pmatrix} \mathbf{v}_i\left(t_n - \frac{h}{2}\right) \\ \vdots \\ \mathbf{r}_i(t_n) \end{pmatrix}$$

$\det J = 1$

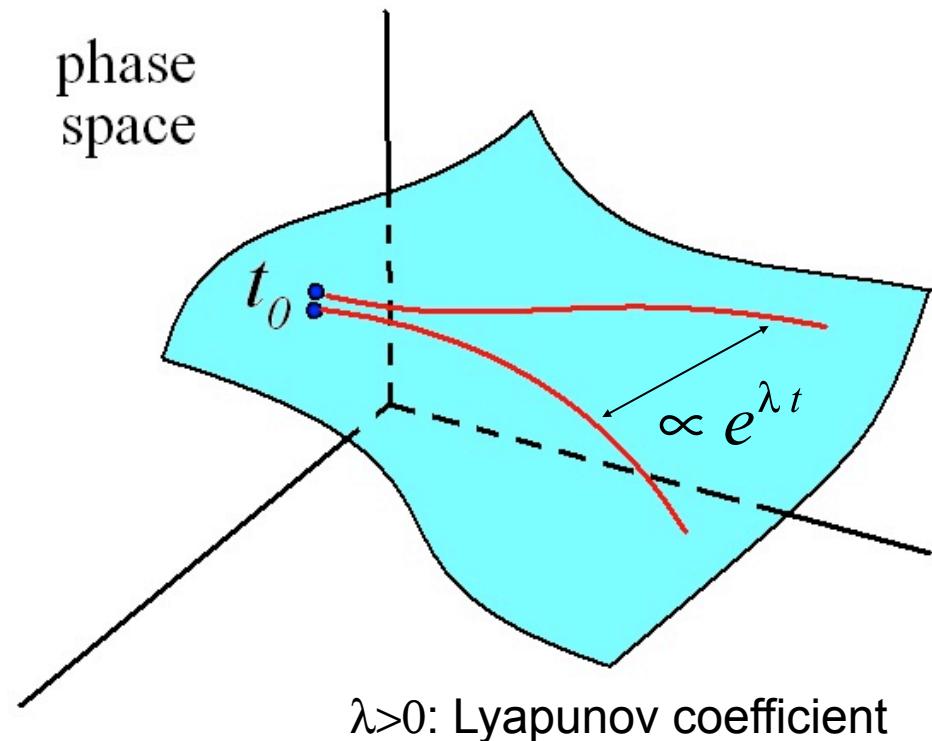
SYMPLECTIC
ALGORITHM

Space volume is conserved (as required by Liouville's theorem of Statistical Mechanics)

Classical many-body systems
are intrinsically CHAOTIC

Therefore the algorithm
cannot reproduce trajectories
exactly

Only requirements are
reversibility and *symplecticity*



Thermodynamic, dynamic, microscopic properties result as TIME AVERAGES over phase space trajectories. For example, the energy:

$$E = \langle H \rangle = \frac{1}{\tau} \sum_n H_n$$

Extensions of constant-energy MD:

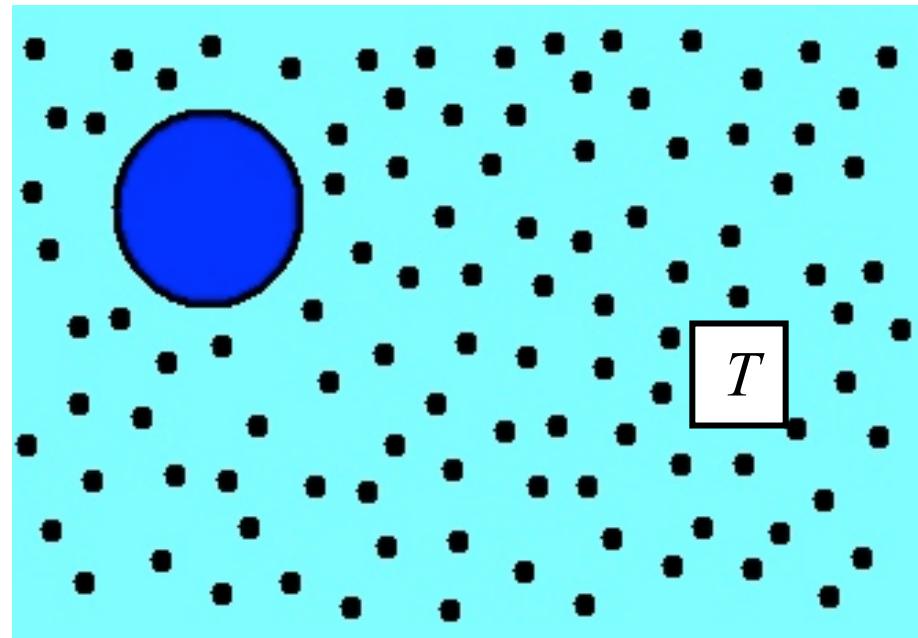
- **Isothermal MD:**

the system is coupled to an external heat-bath (thermostat at specific temperature T), with which energy is exchanged to maintain a constant temperature

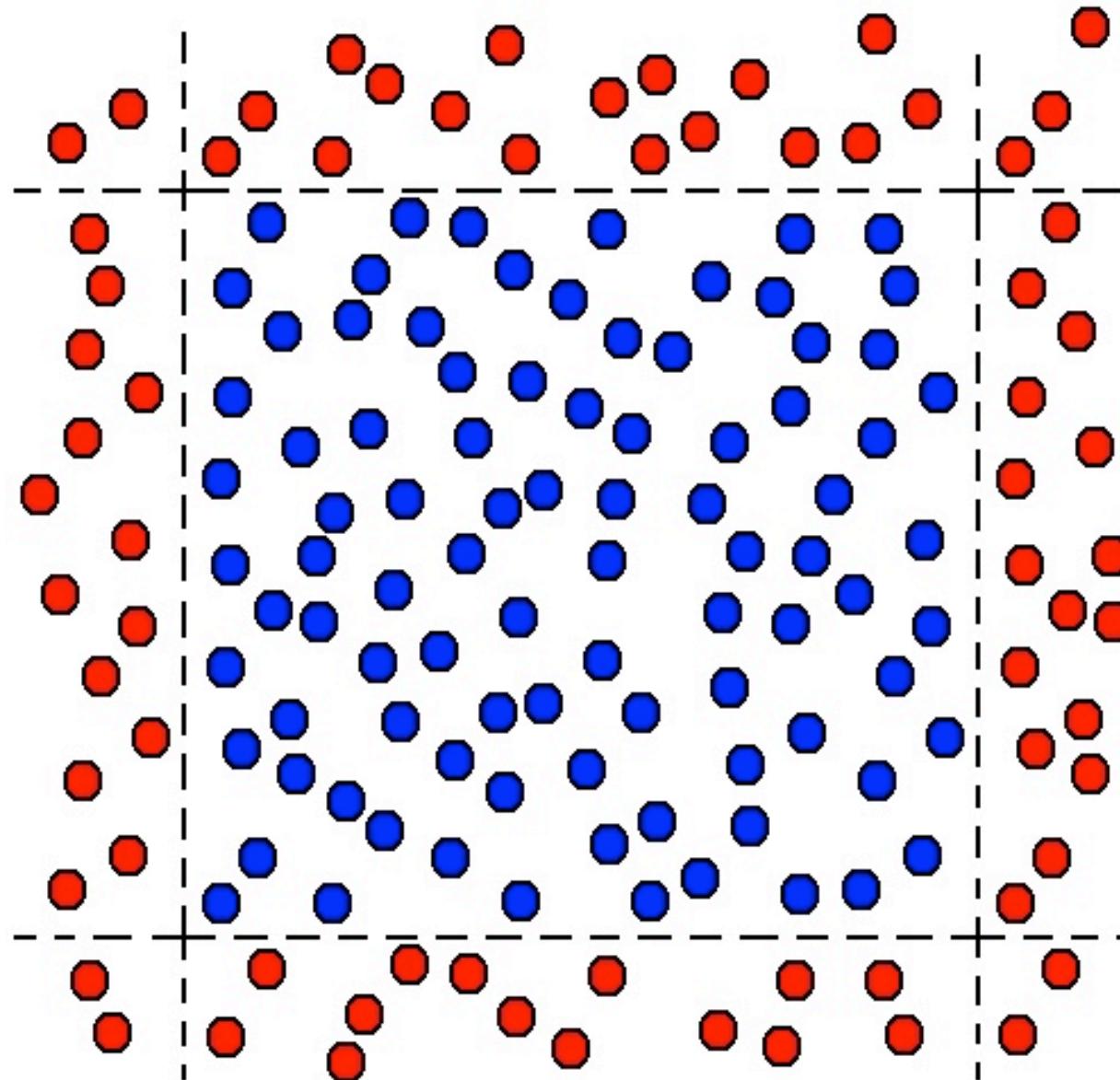
Example: Langevin thermostat

$$m_i \frac{d\mathbf{v}_i}{dt} = -\xi \mathbf{v}_i + \mathbf{g}_i + \mathbf{F}_i$$

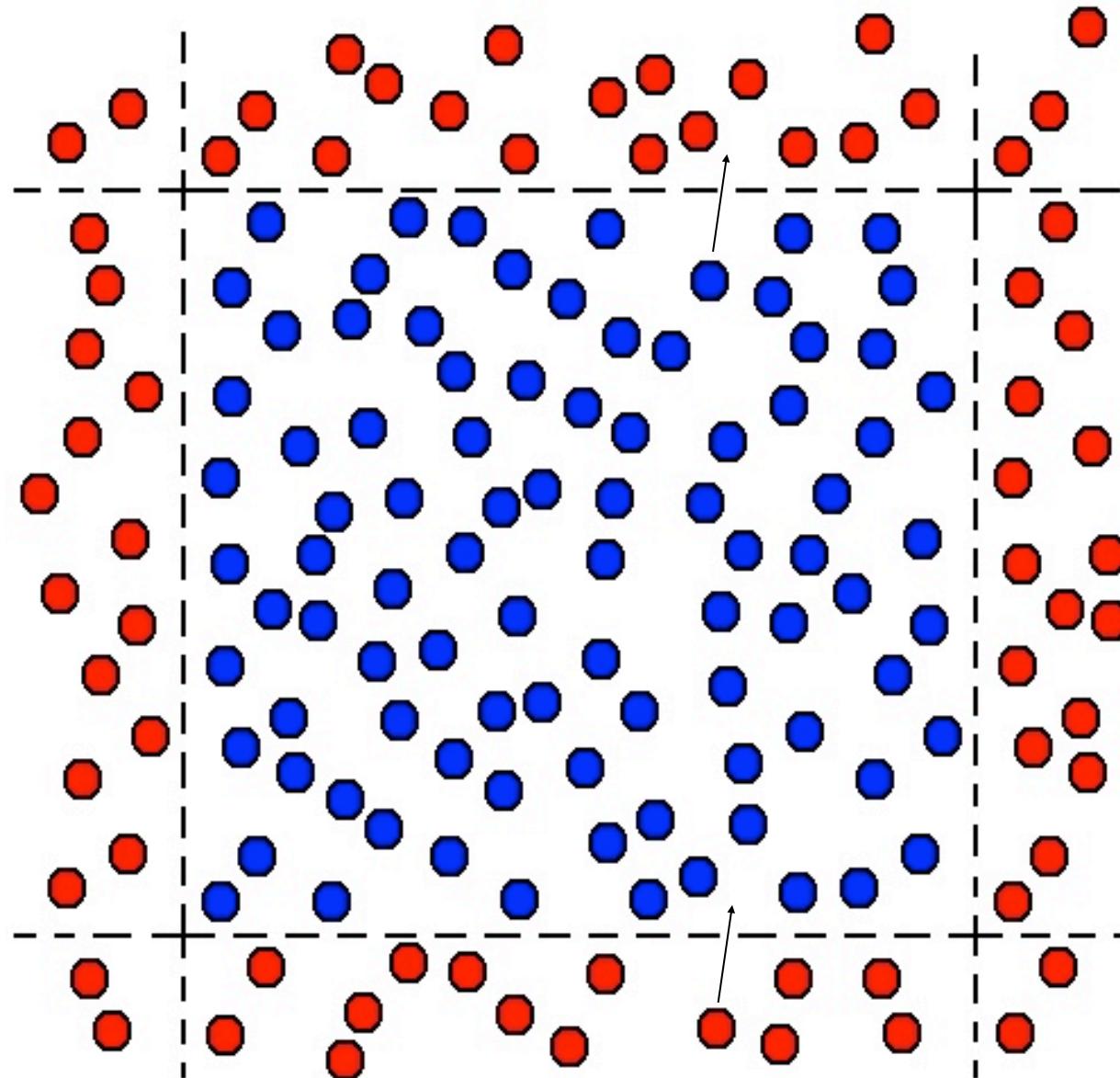
random force
friction term
deterministic force



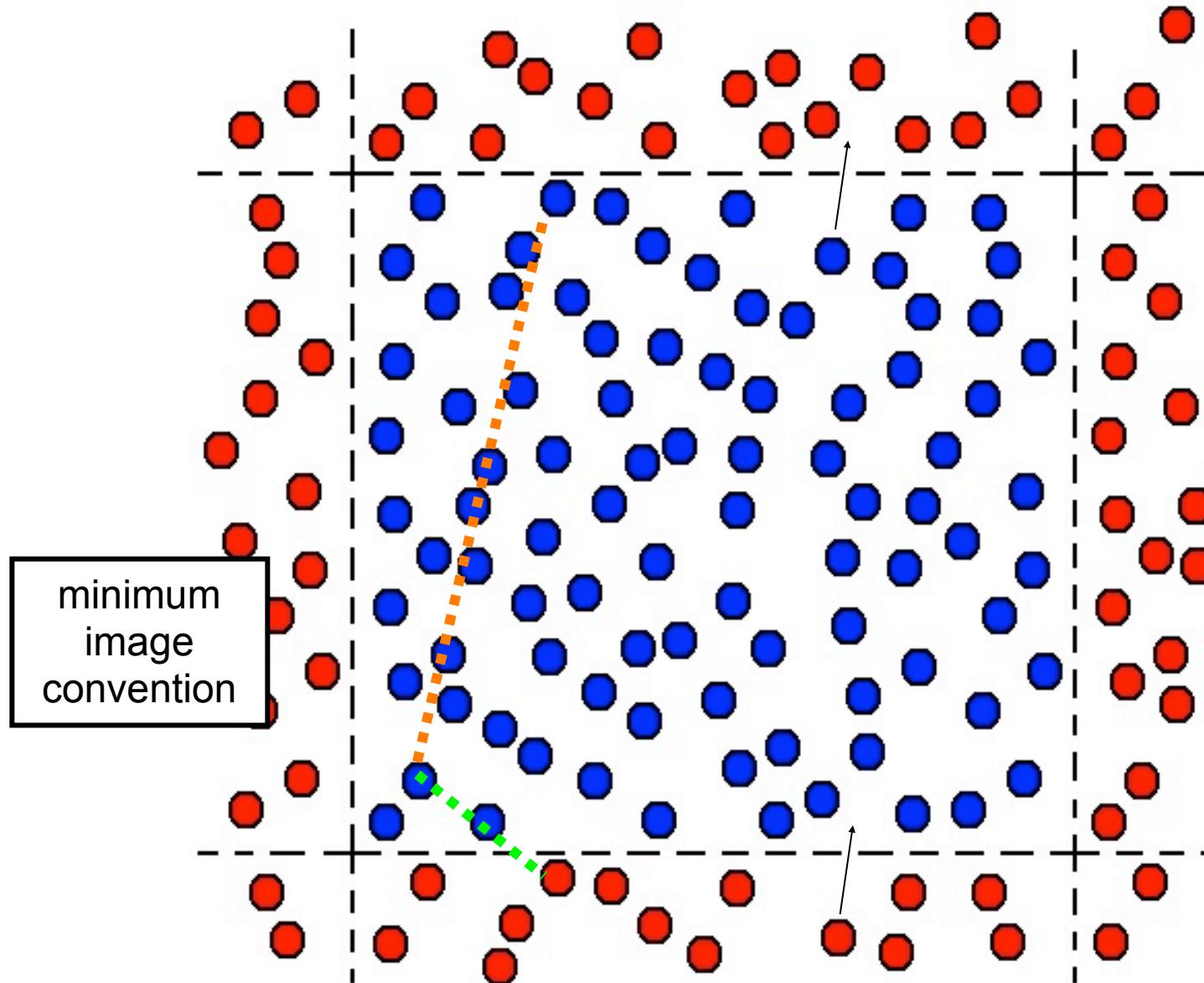
periodic boundary conditions (2D)



periodic boundary conditions (2D)



periodic boundary conditions (2D)



The time-consuming part of the calculation is the force:

$$m_i \frac{d\mathbf{v}_i}{dt} = \mathbf{F}_i$$

assuming two-body forces



$$\mathbf{F}_i = \sum_{j \neq i} \mathbf{F}_{ij}$$

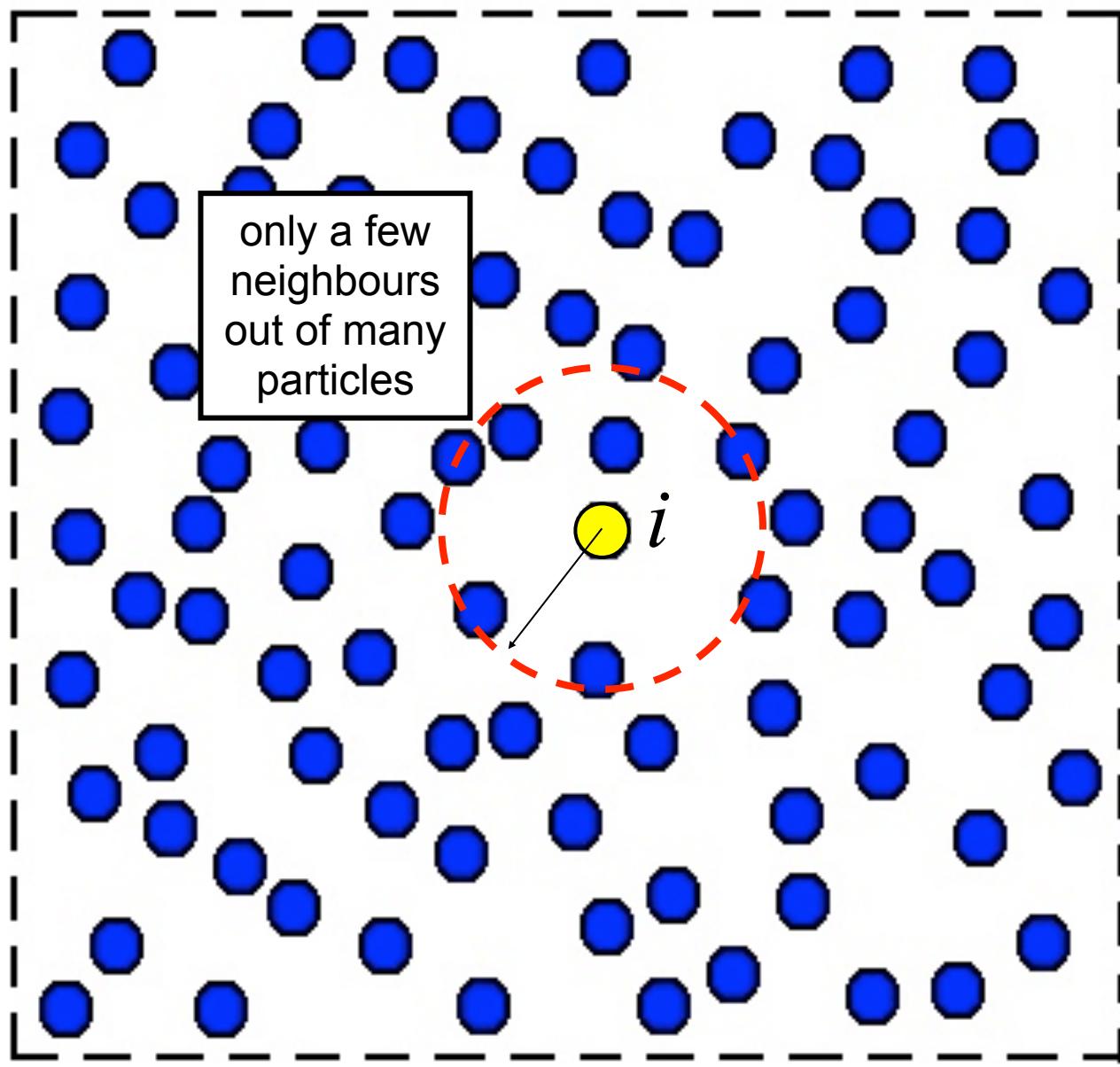
Force calculation amounts to up to 90% of
CPU time required in whole calculation

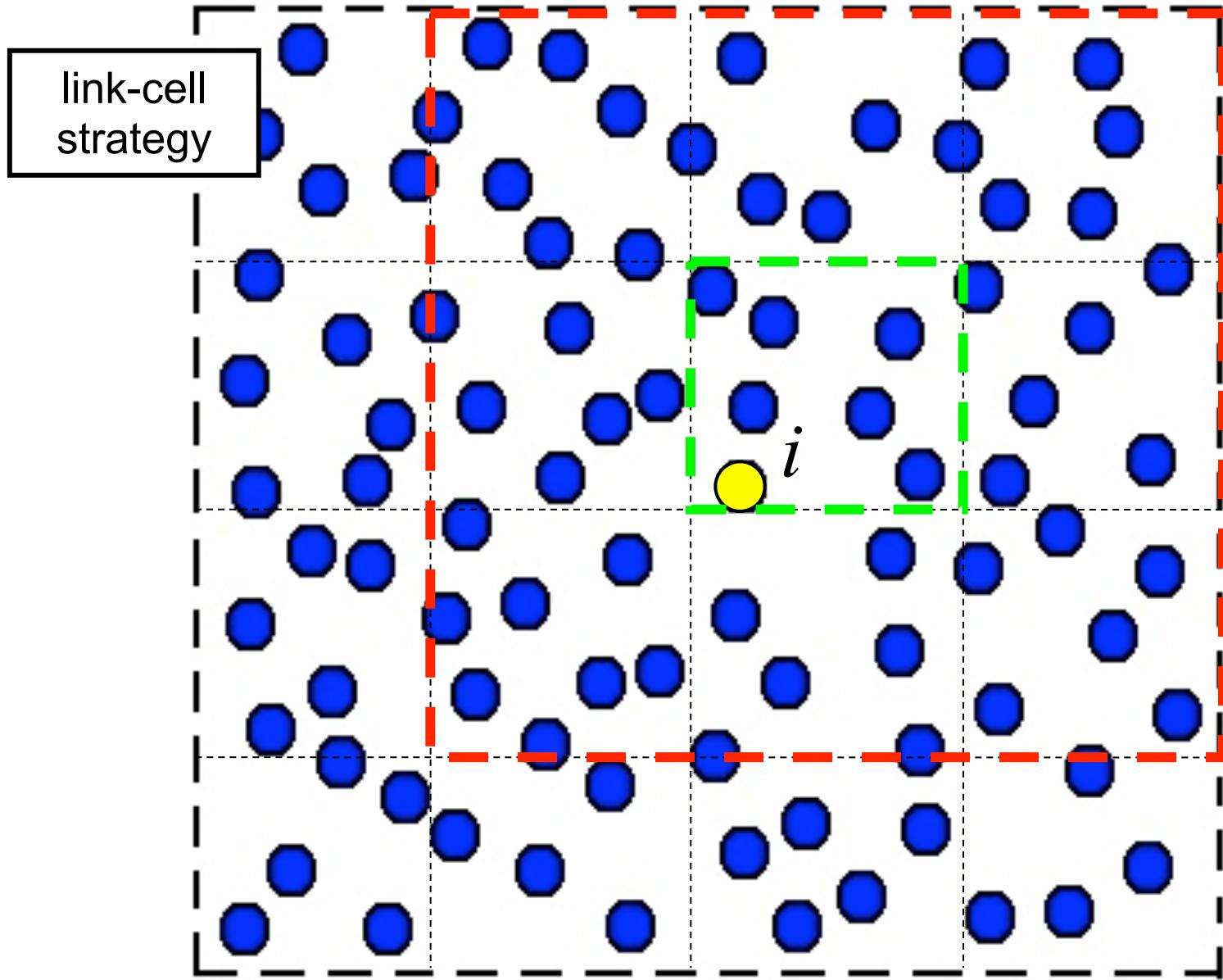
➡ mayor barrier to large-scale simulations
need for new algorithms and techniques

The sum is over all
neighbours within
the interaction
distance. But,
which are they?

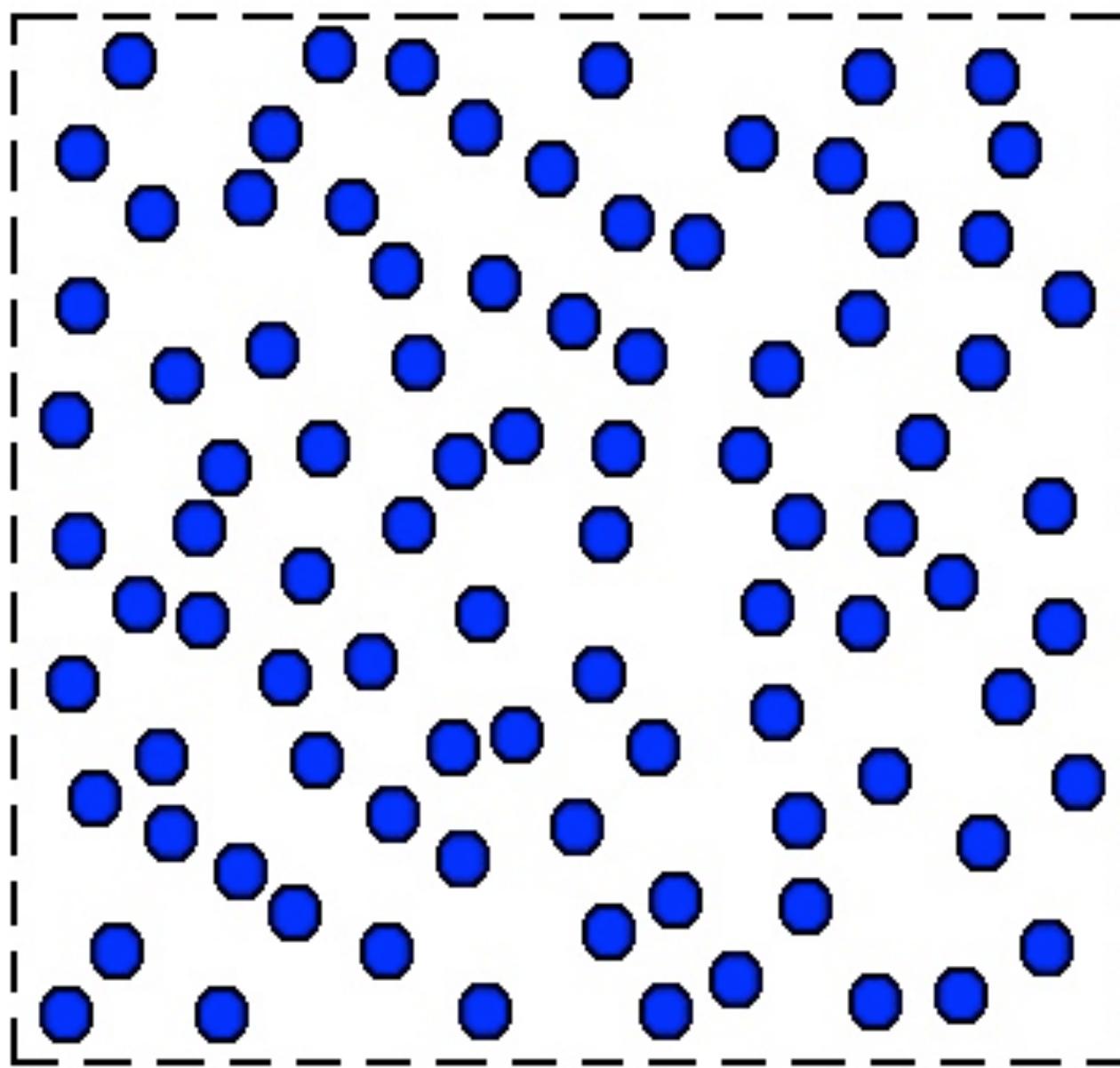


Identify all
distinct pairs

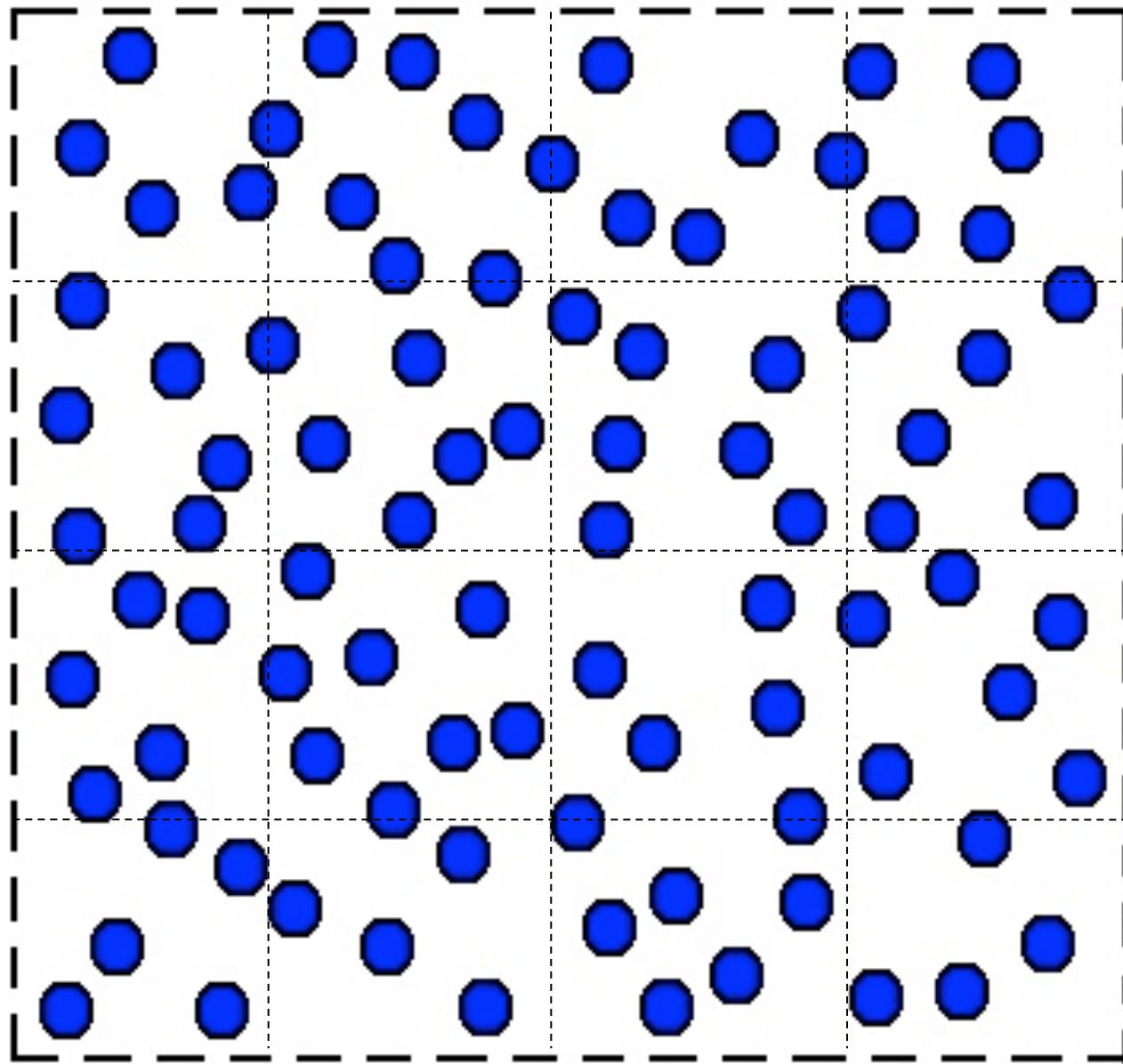




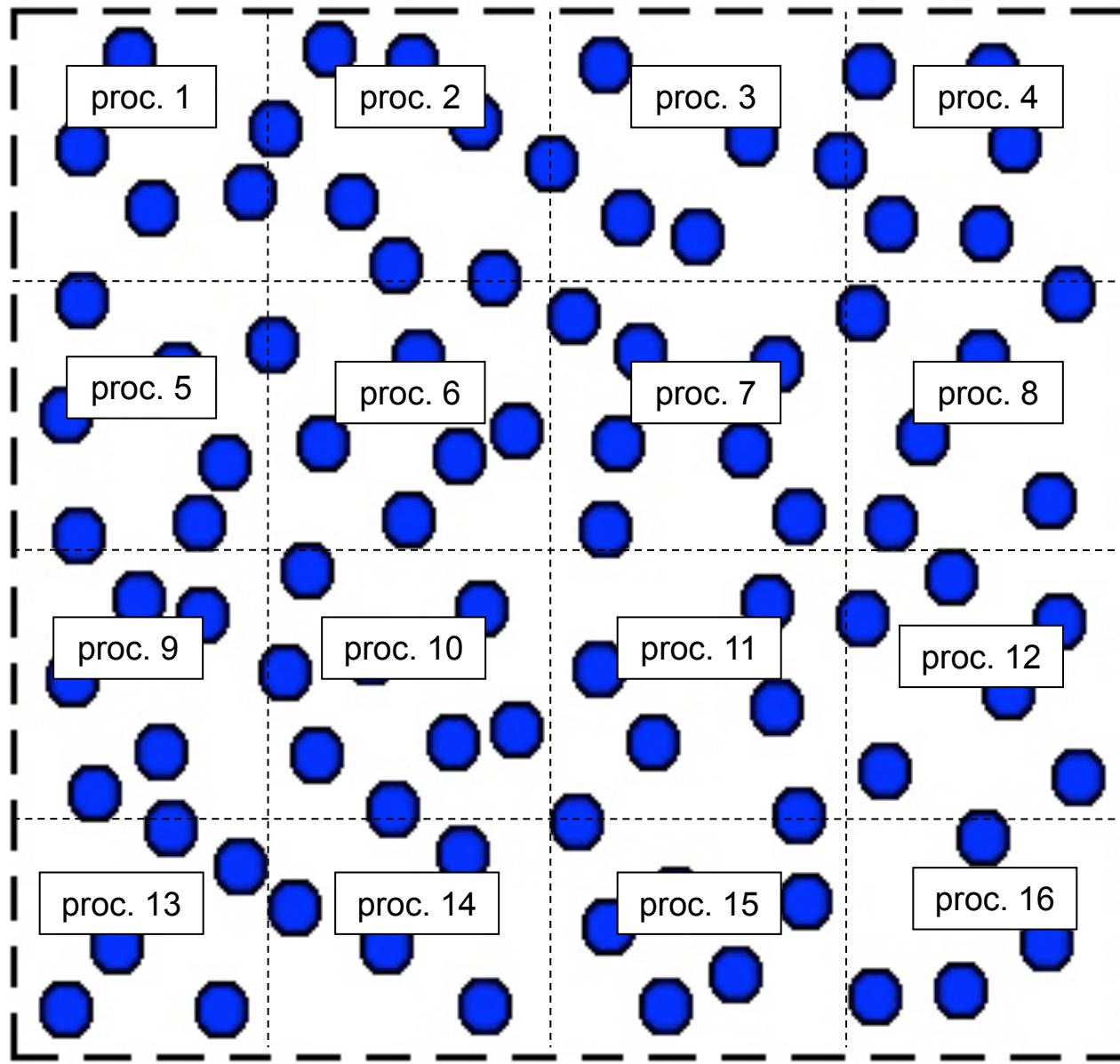
parallelisation: domain decomposition



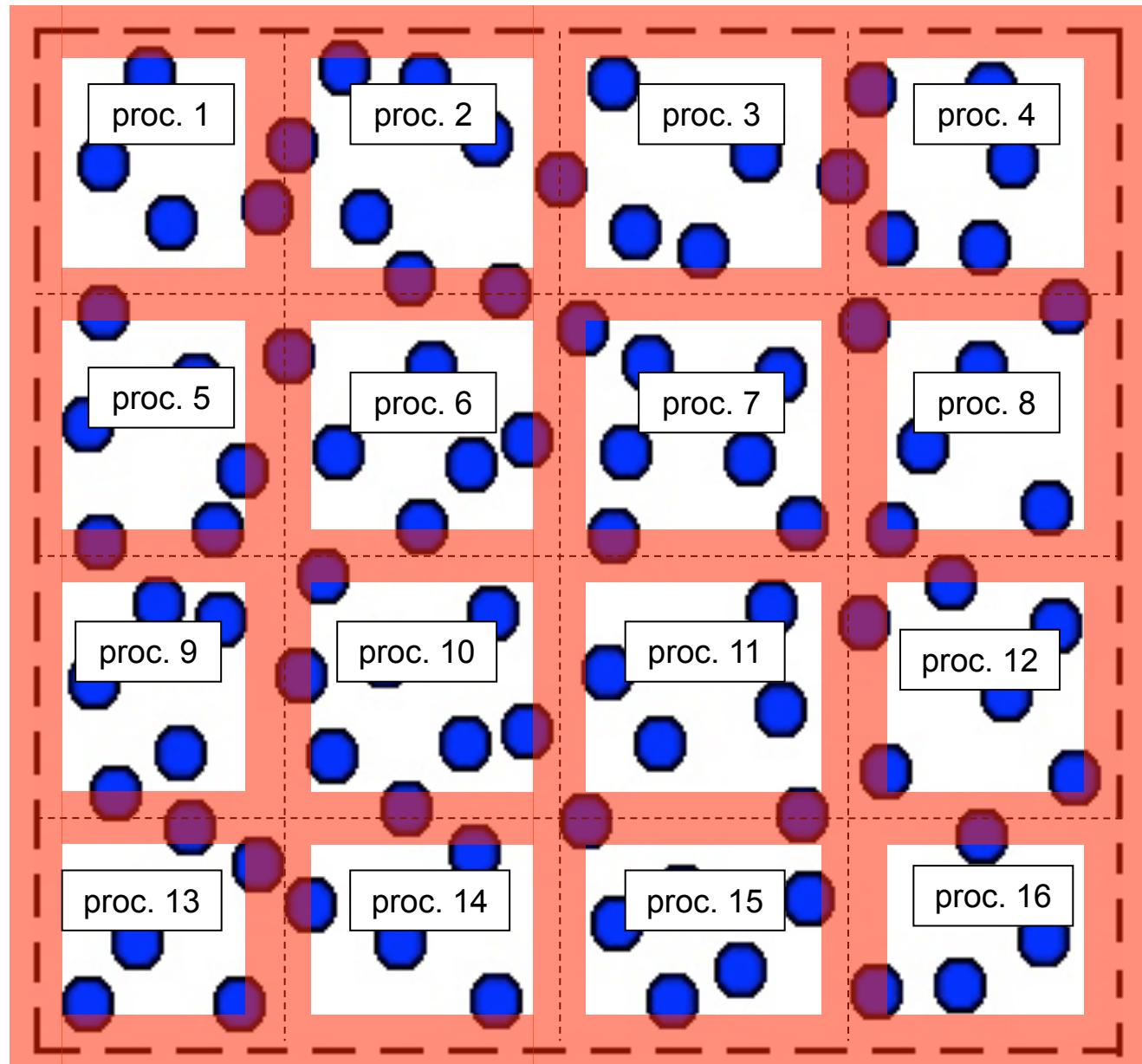
parallelisation: domain decomposition



parallelisation: domain decomposition



parallelisation: domain decomposition



- **GPU (Graphics Processing Unit)**

Specialised processor for graphics rendering
(multithreaded, data parallel co-processor)

GPGPU: General purpose computing on GPU



GeForce 8800 GTX (128 cores)



Tesla C1060 (240 cores)

- **CUDA (Compute Unified Device Architecture)**

Arquitectura para la programación en paralelo de la empresa NVIDIA

Permite acceder a la GPU para calcular como si fuera una CPU

Programables inicialmente en OpenGL (orientado a gráficos)

CUDA C: lenguaje C extendido

Actualmente también admite códigos en C++, FORTRAN, Java...

Página web de CUDA en NVIDIA:

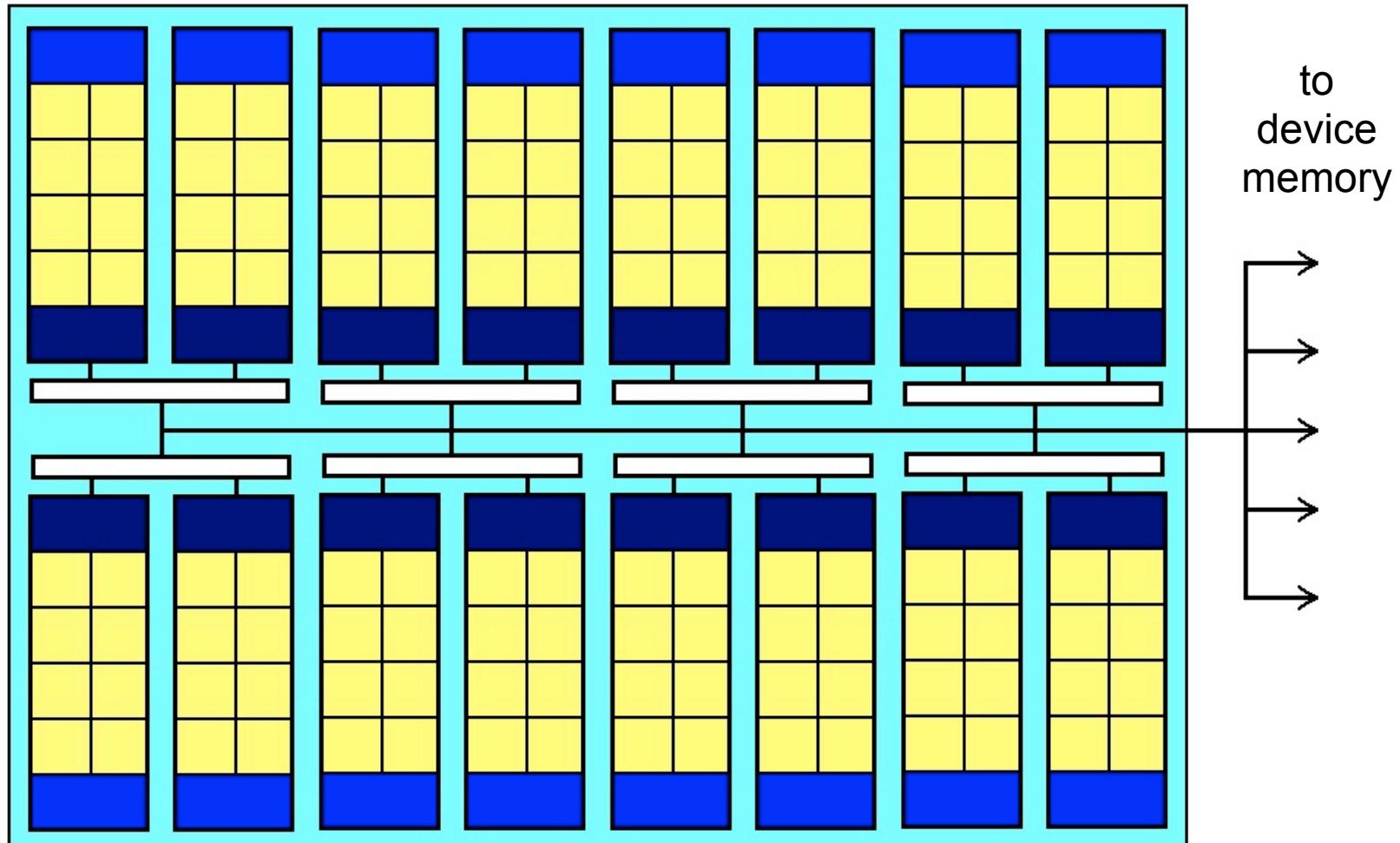
http://www.nvidia.com/object/cuda_home.html

Curso de programación en CUDA en UIUC:

<http://courses.ece.illinois.edu/ece498/al>

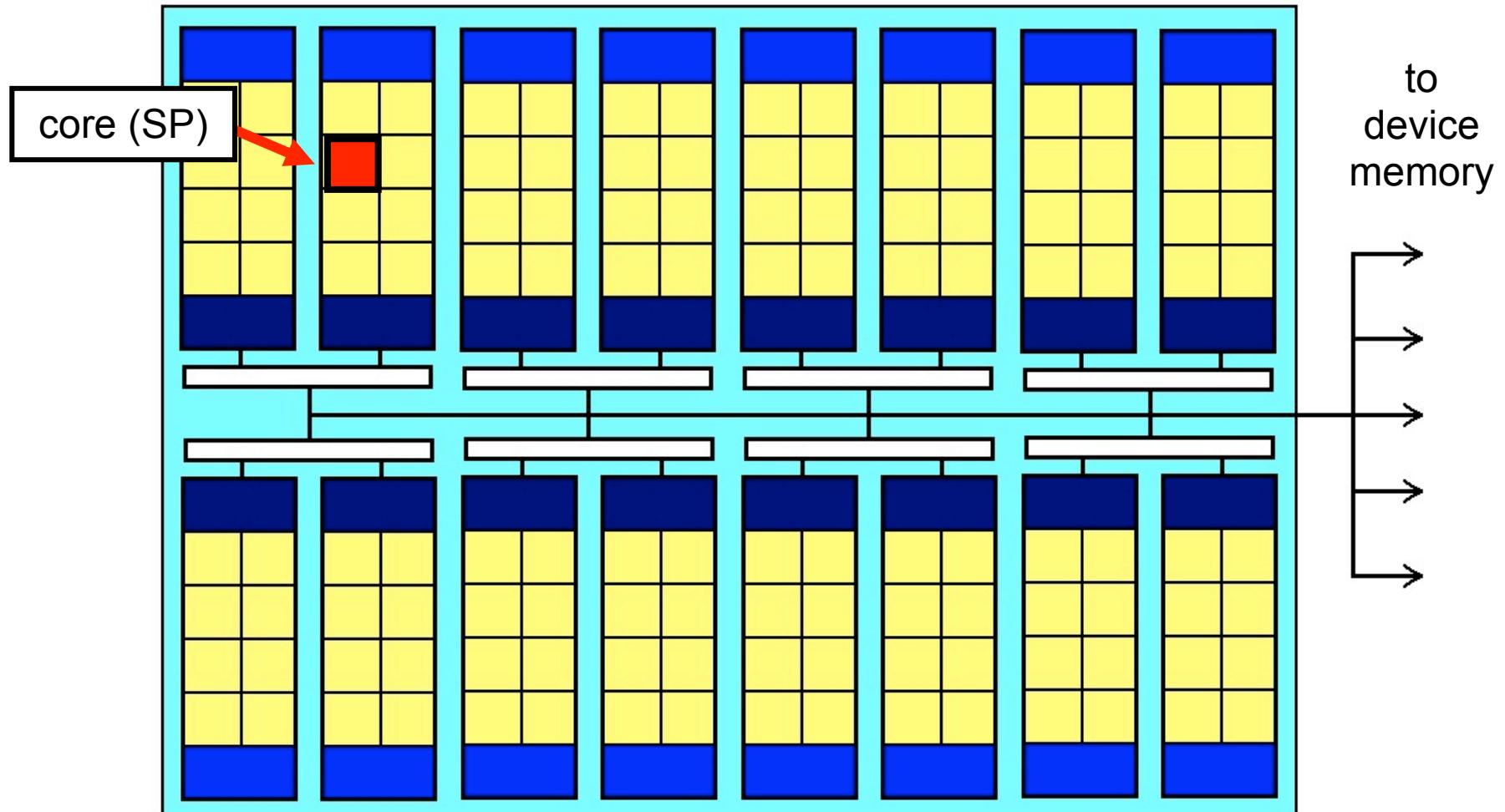
Nvidia GPU design

Tesla
G80

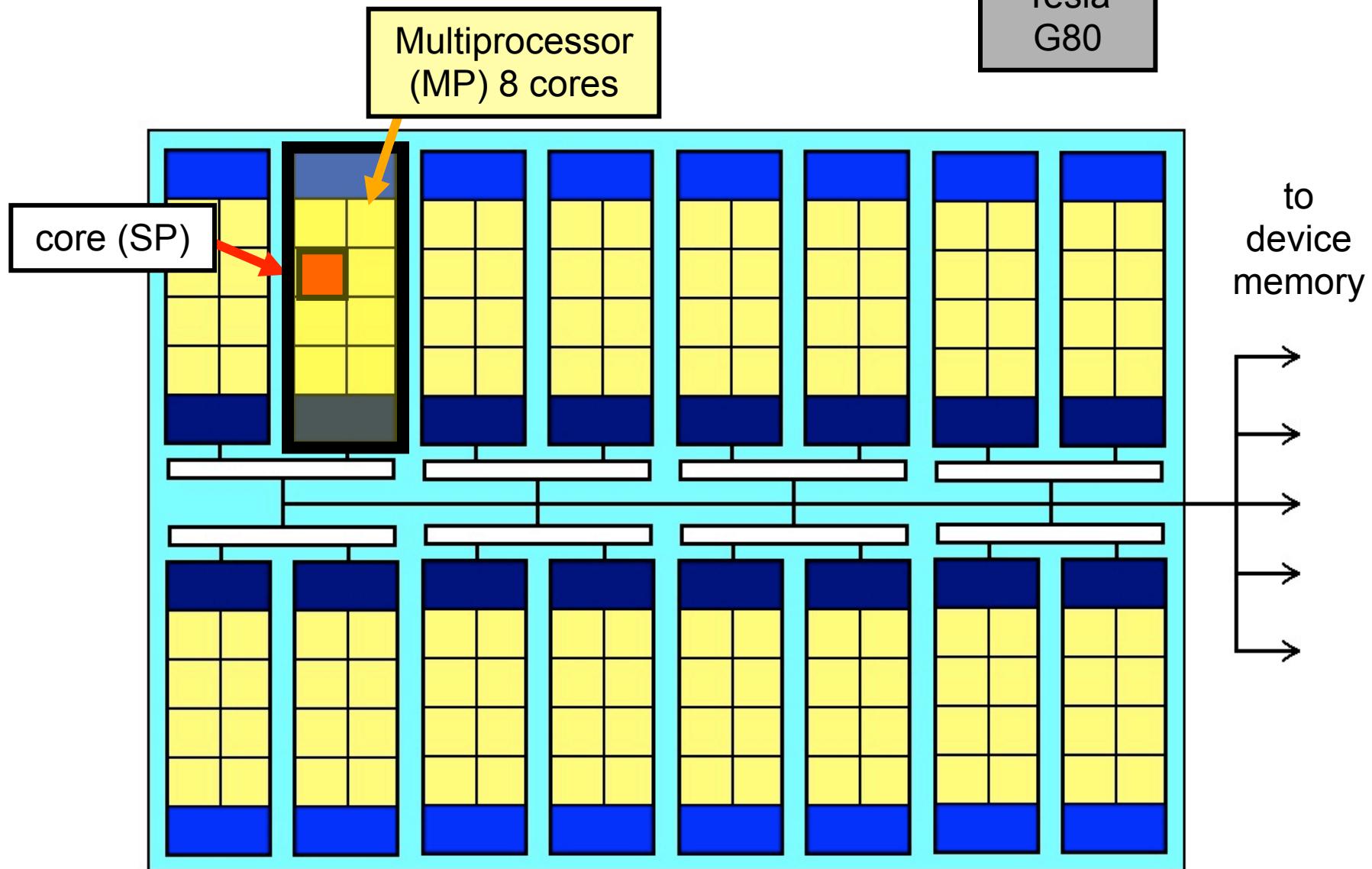


Nvidia GPU design

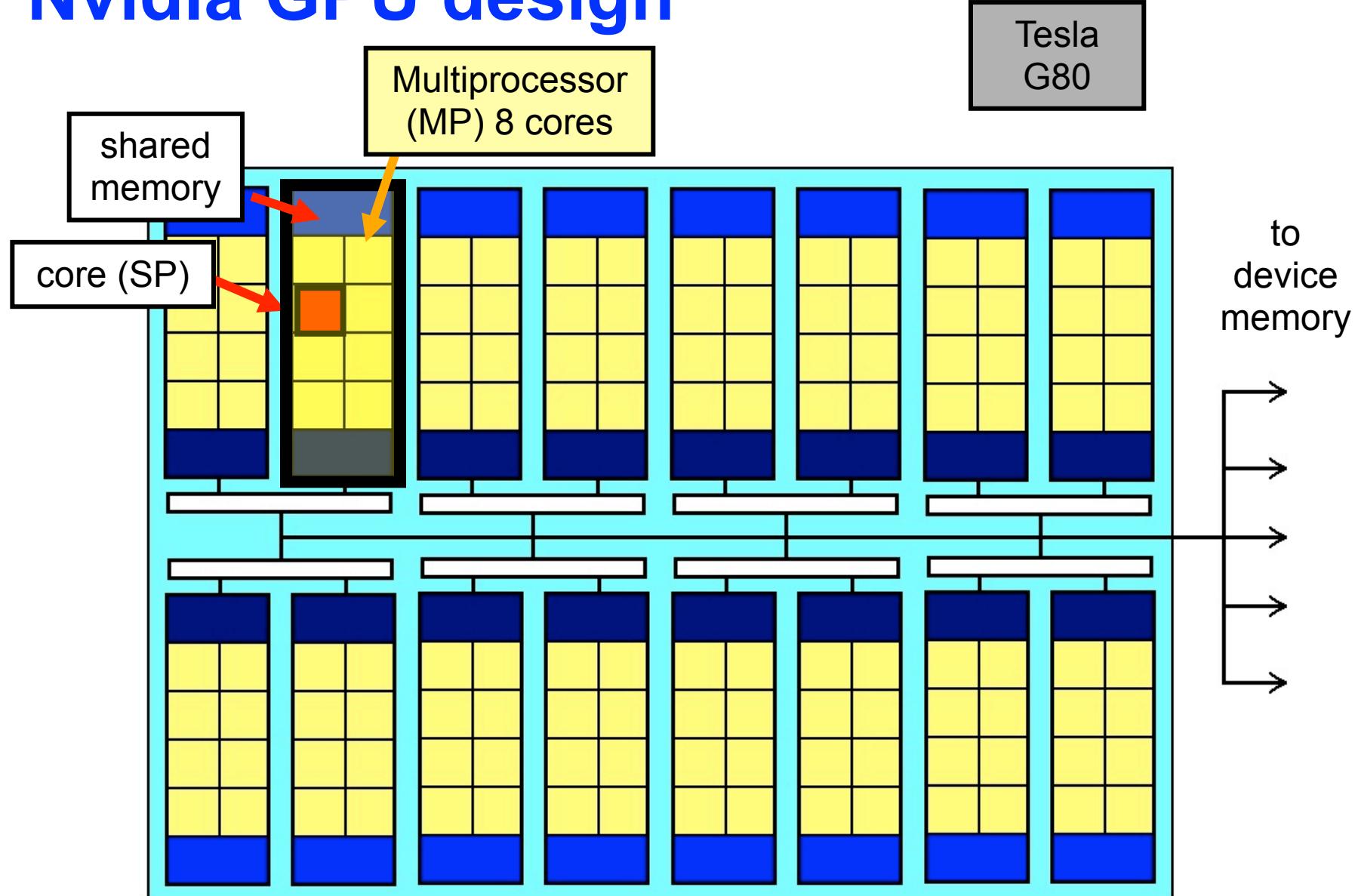
Tesla
G80



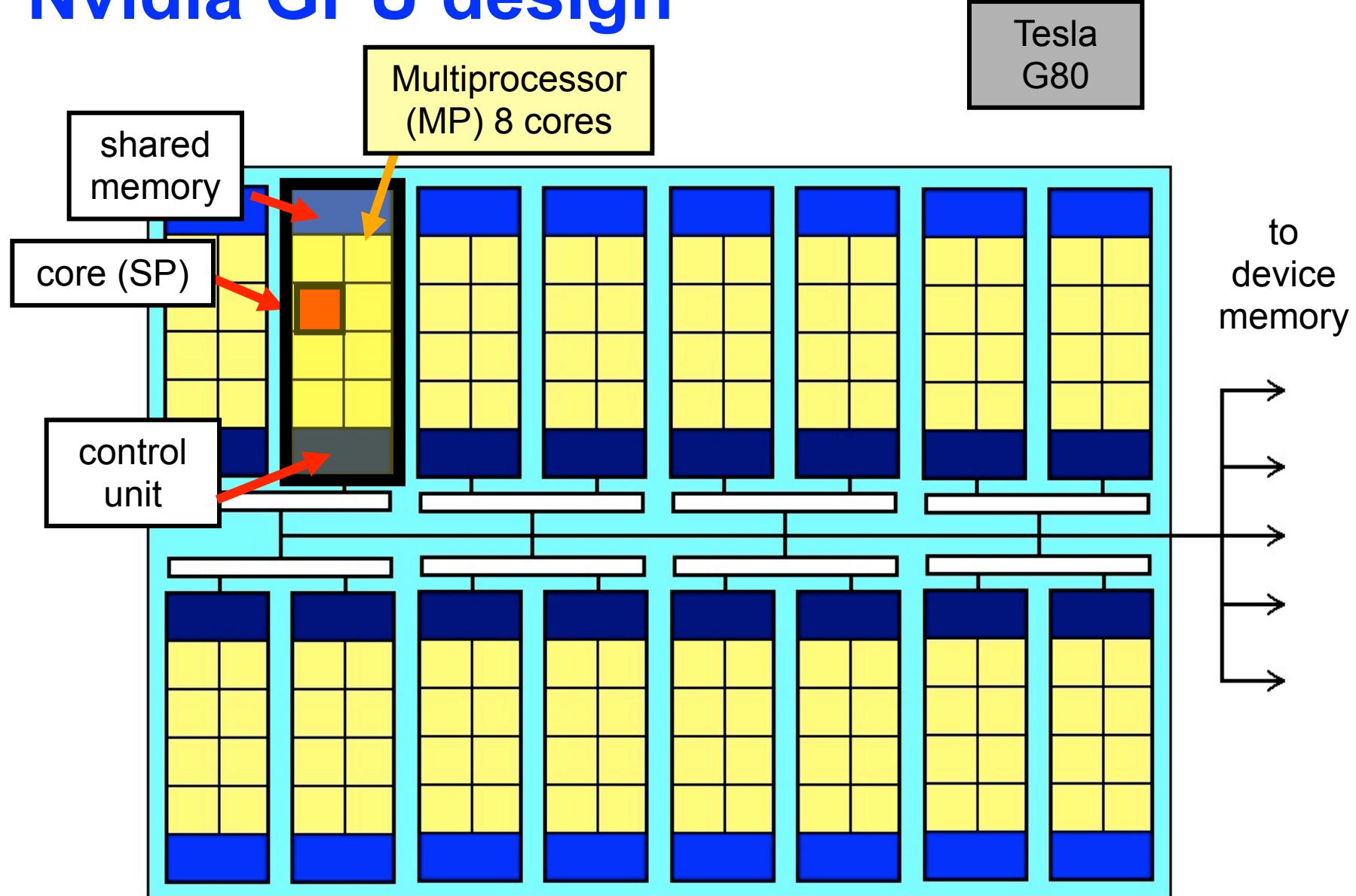
Nvidia GPU design



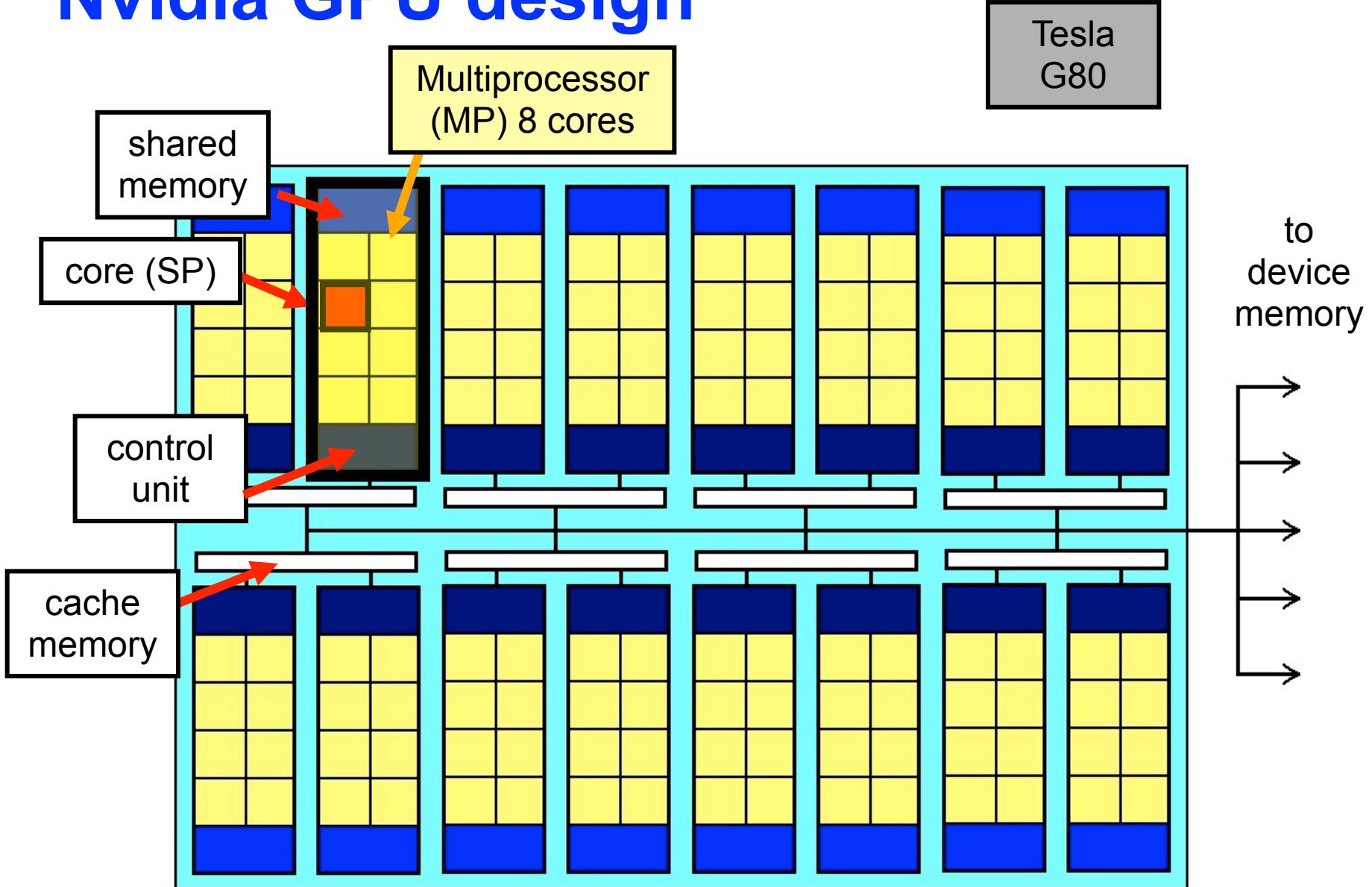
Nvidia GPU design



Nvidia GPU design

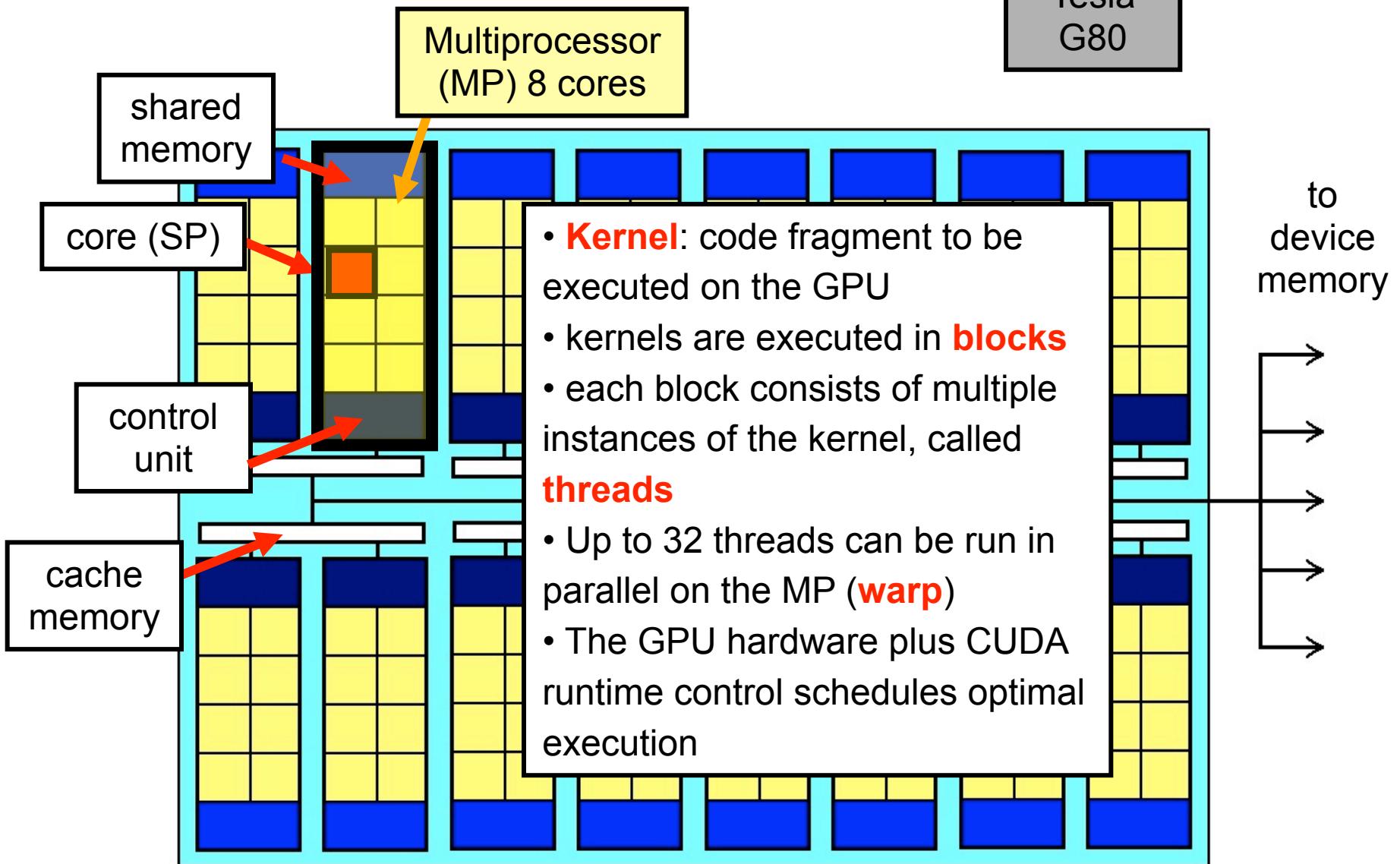


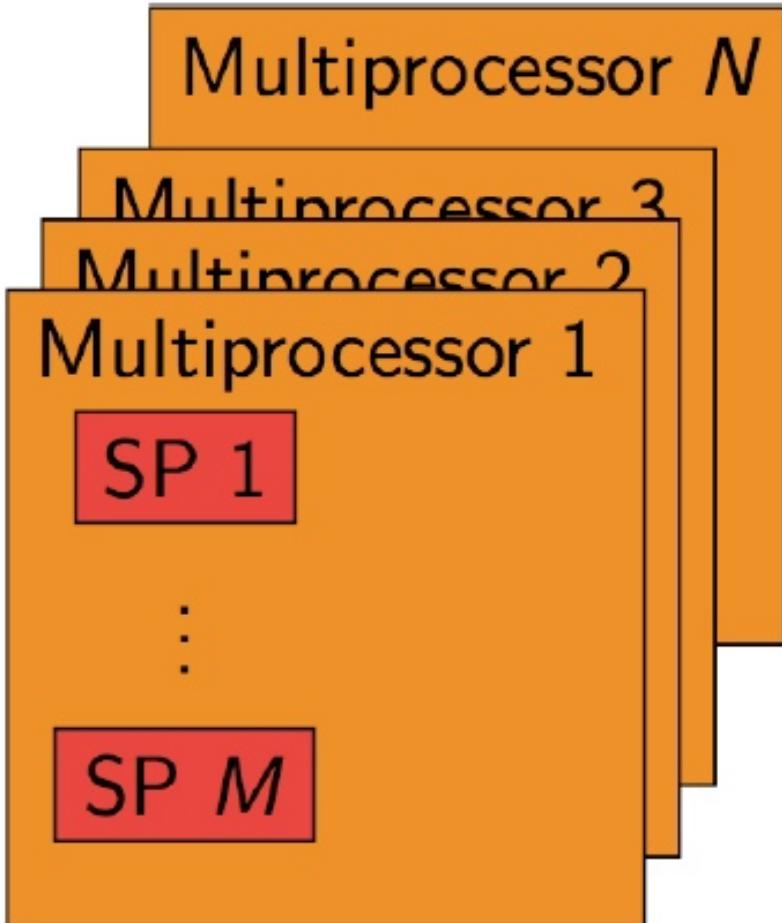
Nvidia GPU design



Nvidia GPU design

Tesla
G80





GeForce GTX580

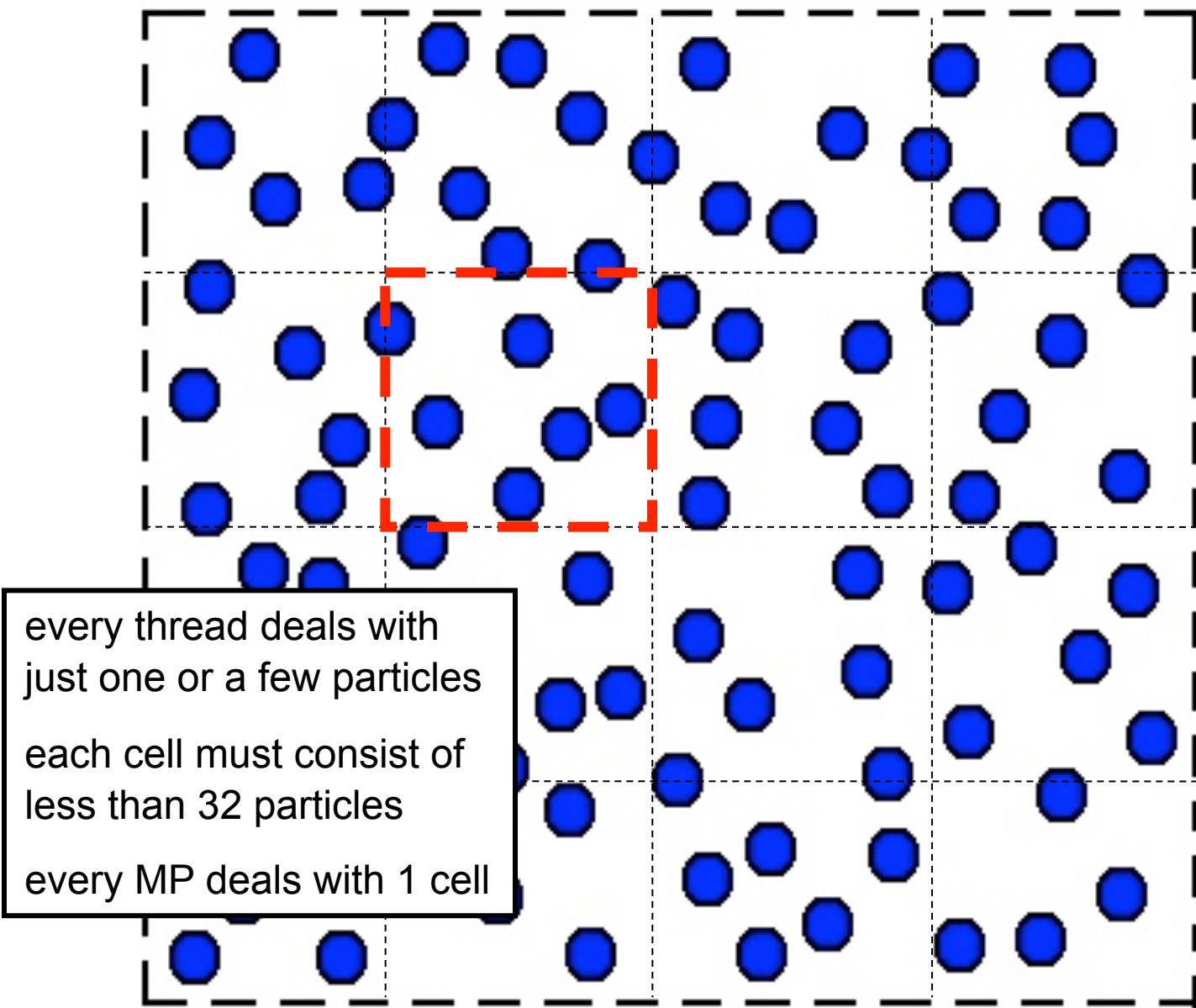
$N = 64$

$M = 8$

512 CUDA SP cores

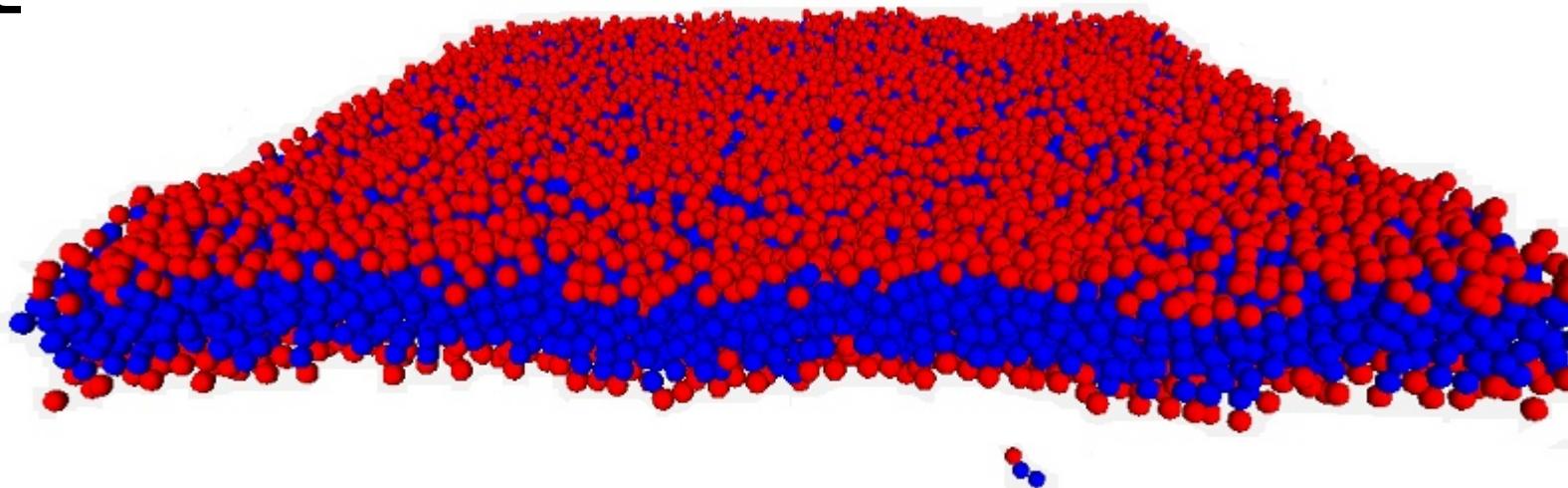
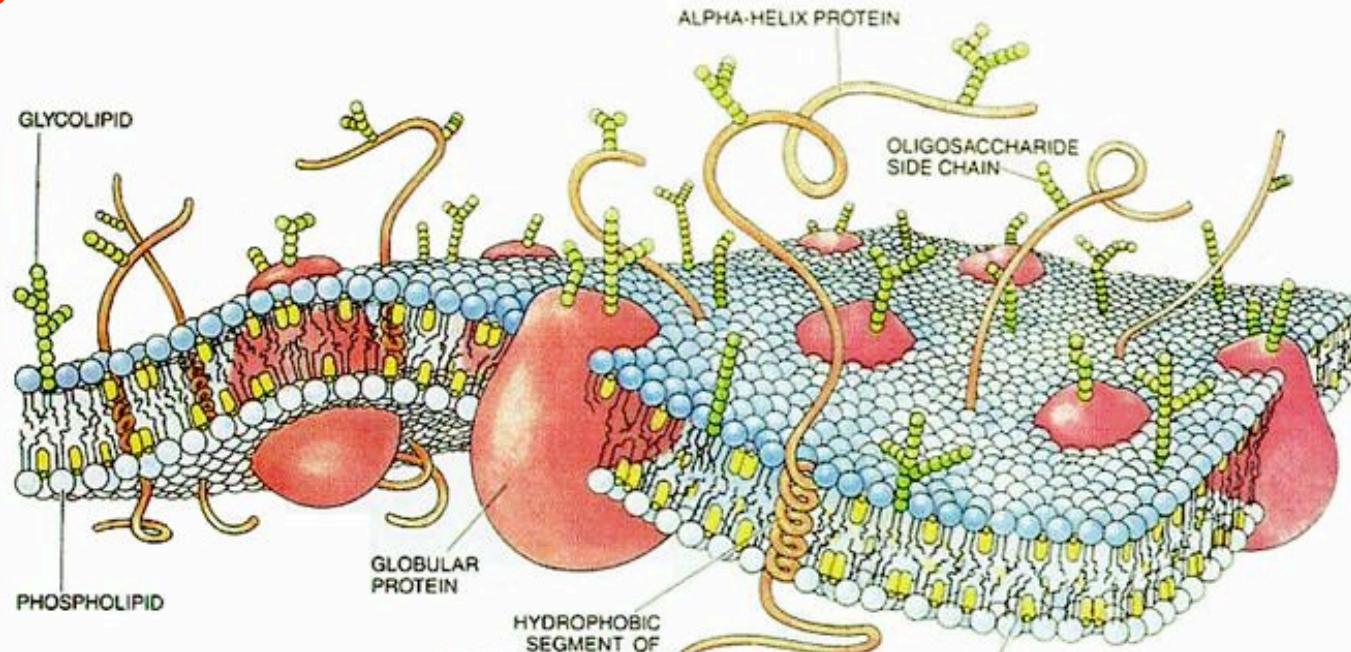
64k shared memory

Teraflop performance
(10^{12} floating point
operations per second)

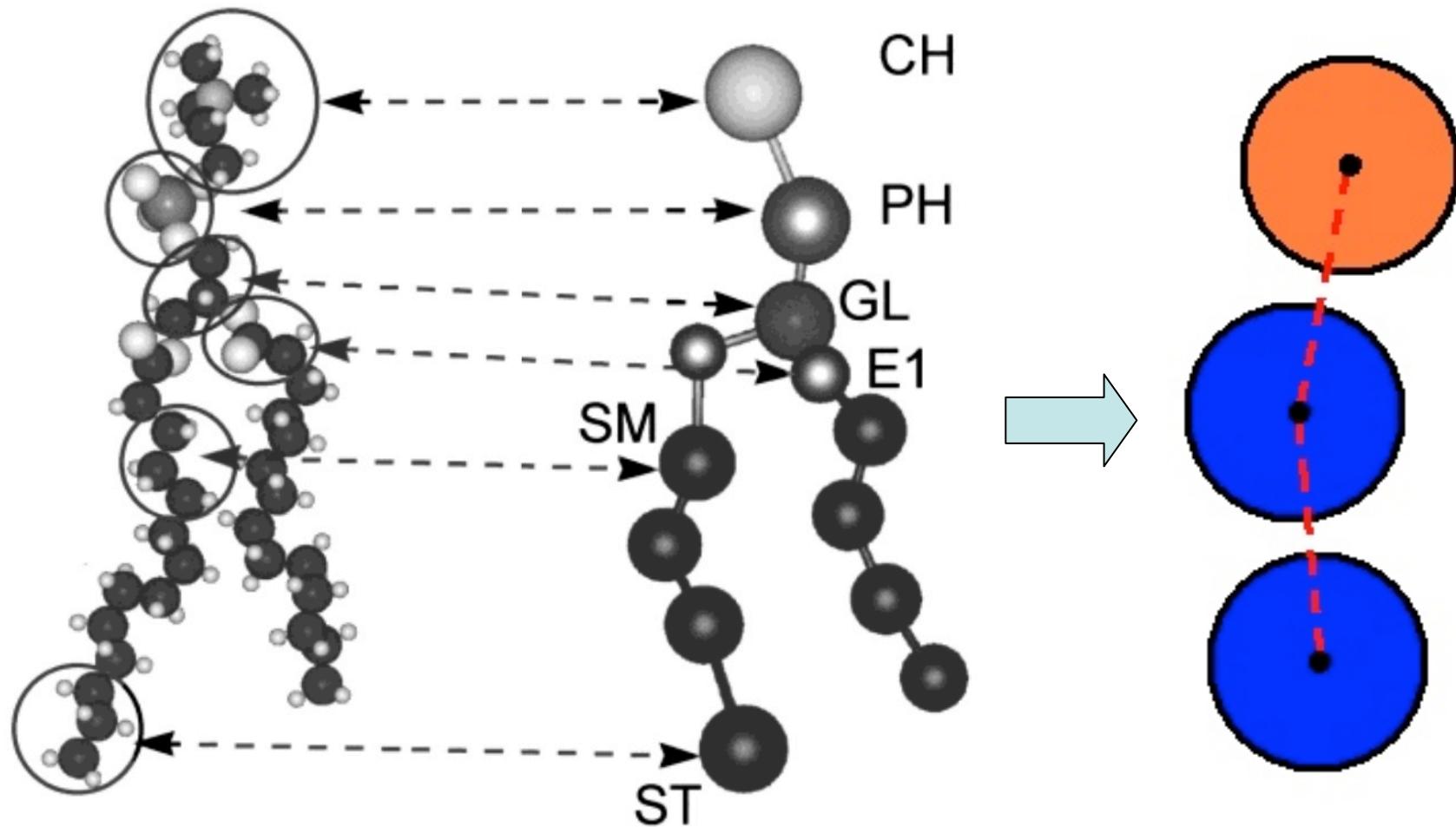


Biological membrane

mesoscopic model

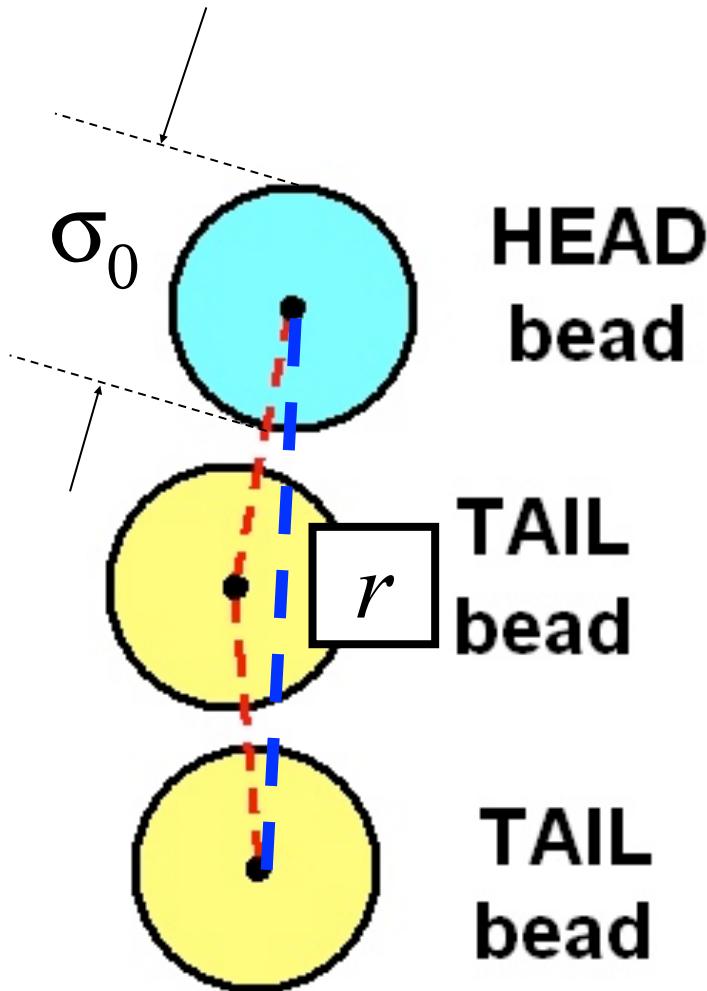


Coarse-grained molecular model



Molecular model

Three beads: one head, two tail, no solvent (water)



INTRAMOLECULAR:

- bend

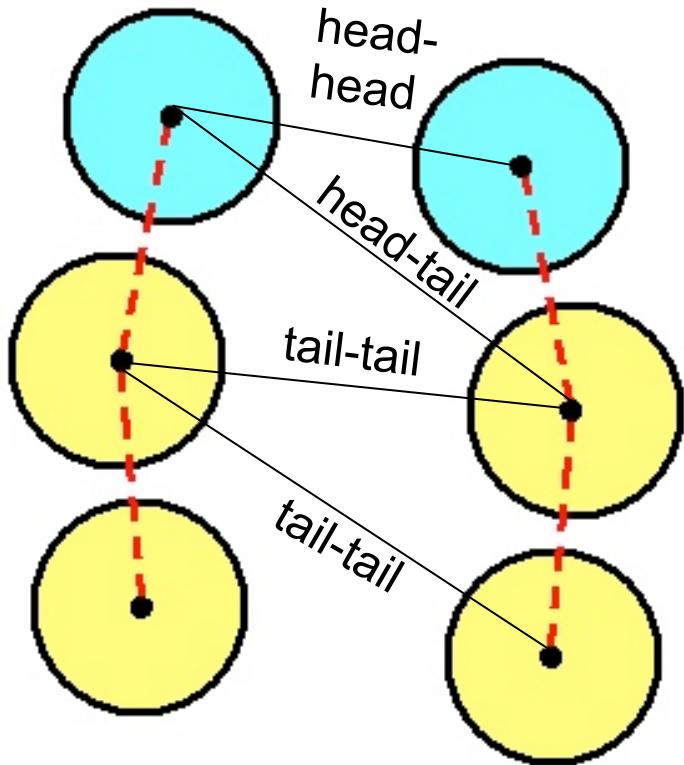
$$V_{bend}(r) = \frac{1}{2} k_{bend} (r - 3\sigma_0)^2$$

$$k_{bend}\sigma_0^2 = 10\varepsilon_0$$

- bond

$$V_{bond}(r) = -\frac{1}{2} k_{bond} r_\infty^2 \log \left[-\left(r / r_\infty \right)^2 \right]$$

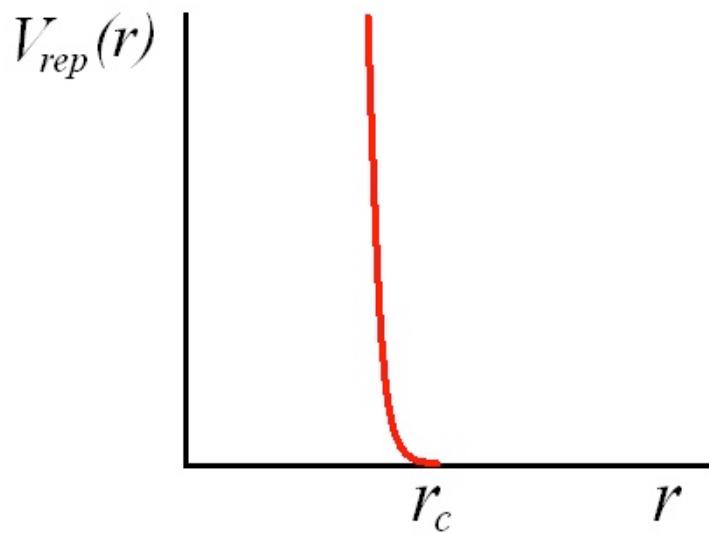
$$k_{bond}\sigma_0^2 = 30\varepsilon_0, \quad r_\infty = 1.5\sigma_0$$



INTERMOLECULAR:

all beads of different molecules interact via the Weeks-Chandler-Andersen potential:

$$V_{rep}(r) = \begin{cases} 4\epsilon_0 \left[\left(\frac{b}{r} \right)^{12} - \left(\frac{b}{r} \right)^6 \right] + \epsilon_0, & r \leq r_c, \\ 0, & r > r_c. \end{cases}$$



ϵ_0 is the unit of energy

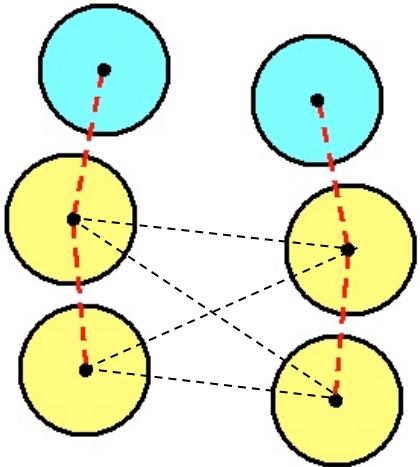
$$b_{head-head} = 0.95 \sigma_0$$

$$b_{head-tail} = 0.95 \sigma_0$$

$$b_{tail-tail} = \sigma_0$$

$$r_c = 2^{1/6} \sigma_0$$

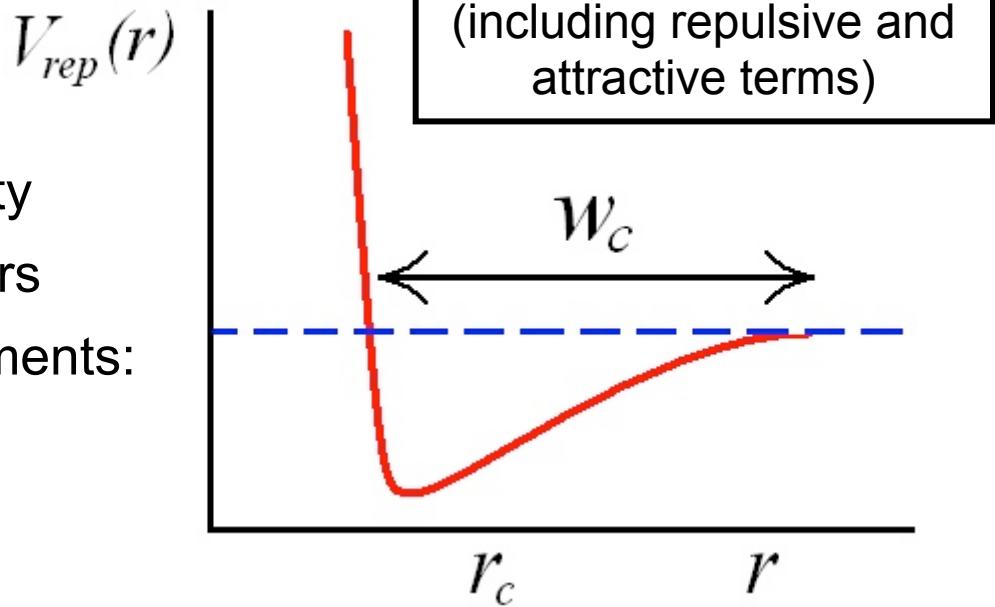
PLUS tail attraction (to mimic hydrophobic effect):



$$V_{attr}(r) = \begin{cases} -\varepsilon, & r < r_c, \\ -\varepsilon_0 \cos^2 \left[\frac{\pi(r - r_c)}{2w_c} \right], & r_c \leq r \leq r_c + w_c, \\ 0, & r > r_c + w_c. \end{cases}$$

Advantages of model:

- Broad range of membrane fluidity
- Easily tunable via few parameters
- Good agreement with measurements: rigidity, diffusion, density



Features of our simulations

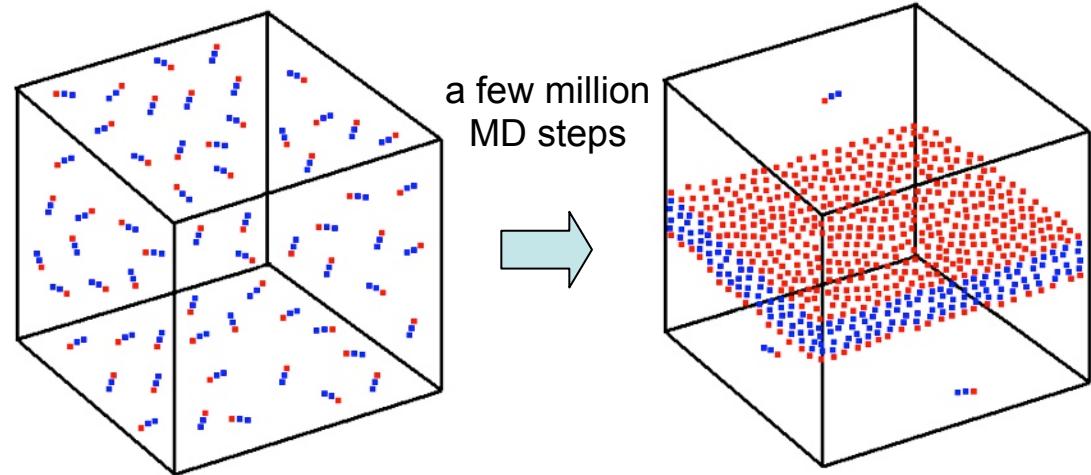
- $N = 7164 - 28800$ molecules ($\times 3$ beads)
- $\tau = 2 - 3$ ns
- NPT ensemble (constant pressure = constant $\gamma = 0$)
(fluctuating system volume)
- Langevin thermostat
- To model typical phospholipid:

$$d = 10 \text{ nm} \text{ (membrane thickness)} \longrightarrow \sigma_0 = 1.7 \text{ nm}$$

$$\varepsilon_0 = 3.72 \times 10^{-21} \text{ J} \quad \rightarrow \quad \tau_0 = \sigma_0 \sqrt{\frac{m_0}{\varepsilon_0}} = 9.85 \text{ ps} \quad \begin{matrix} \text{(time scale of MD} \\ \text{simulation)} \end{matrix}$$
$$m_0 = 220 \text{ g/mol}$$

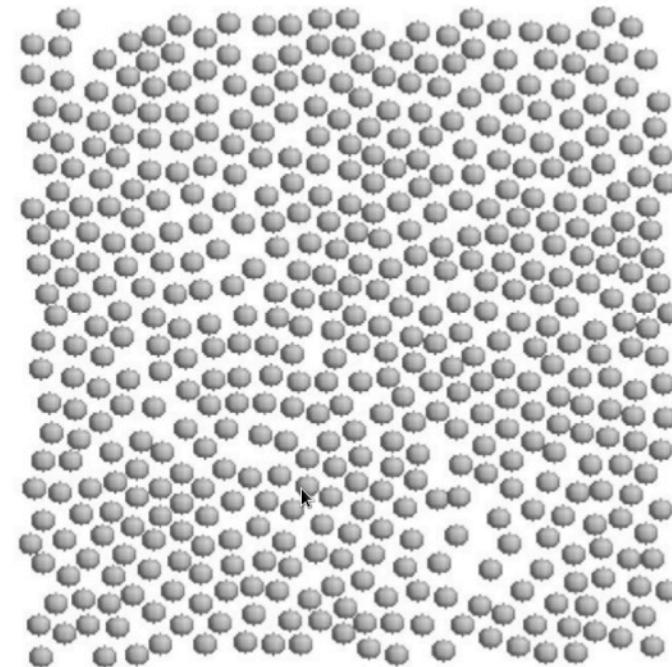
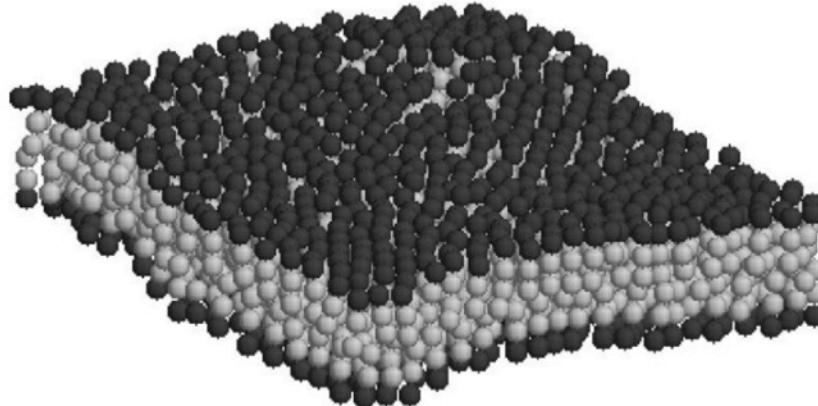
molecules self-assemble spontaneously into planar membranes

special MD techniques
(ensuring $\gamma=0$) are needed
(fluctuating box)



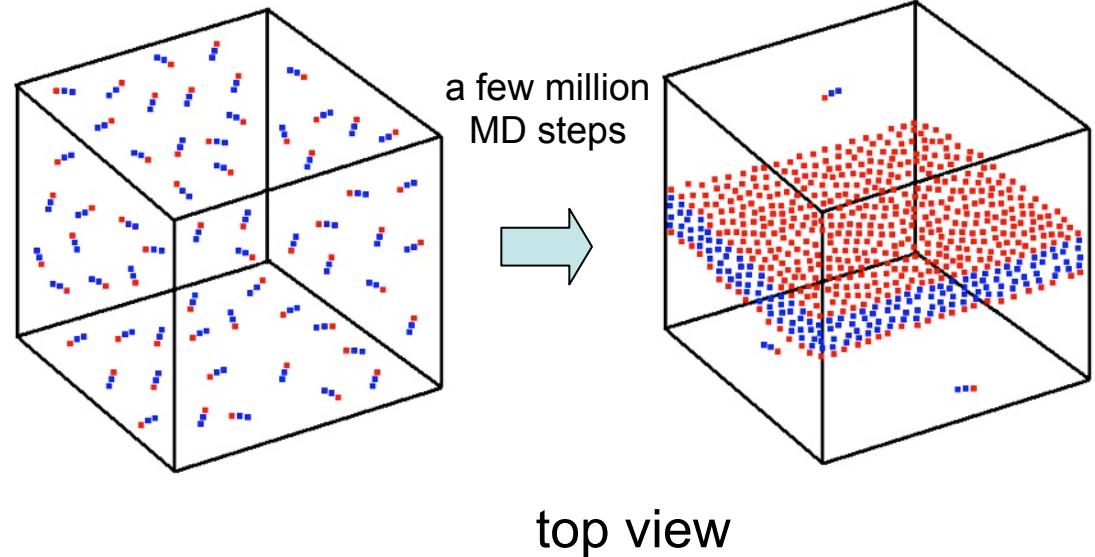
top view

side view



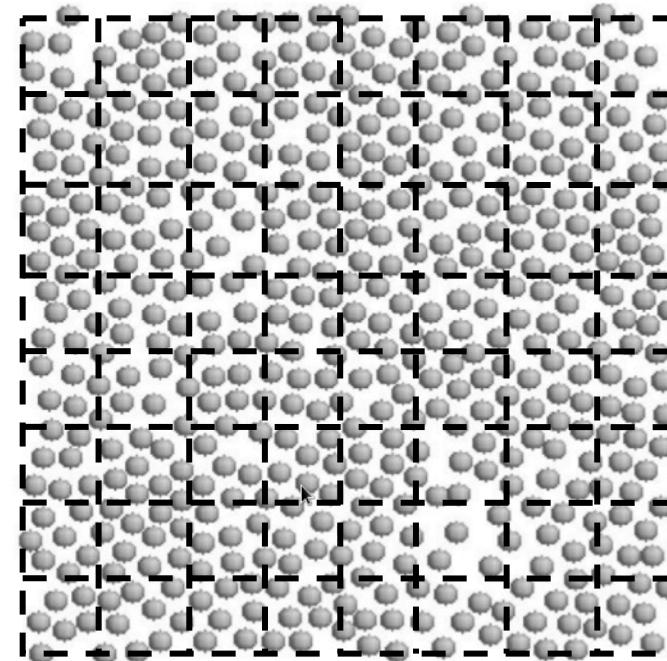
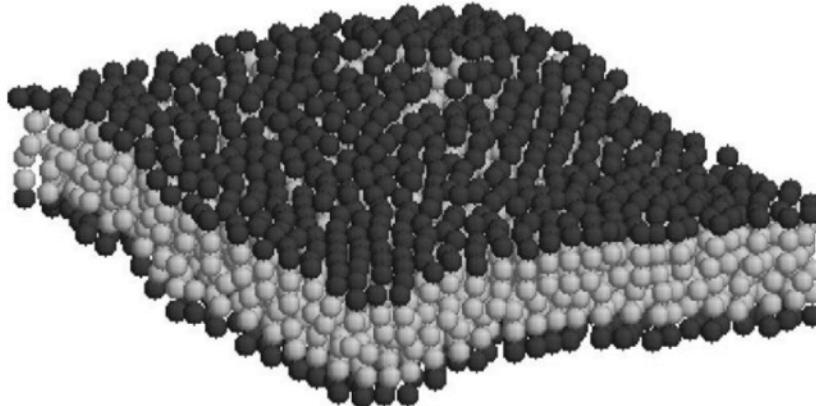
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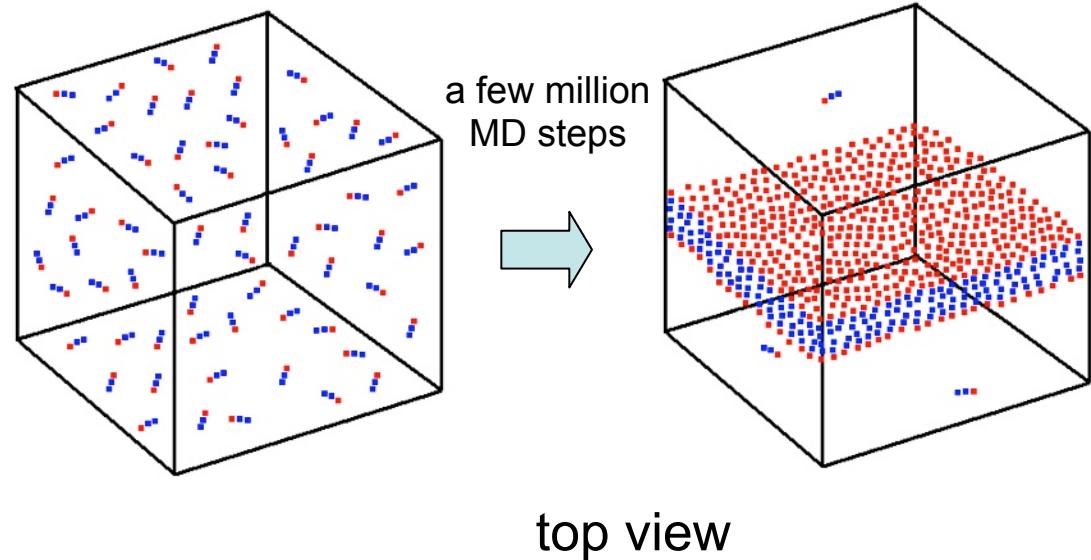
top view

side view



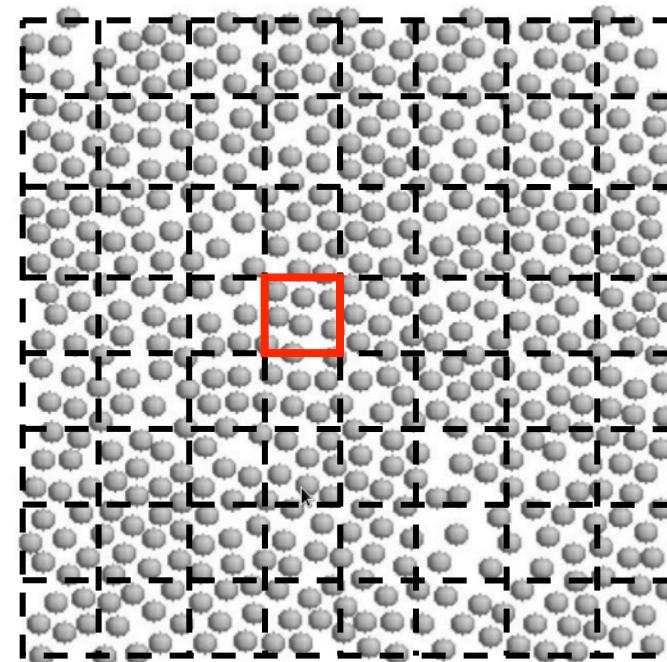
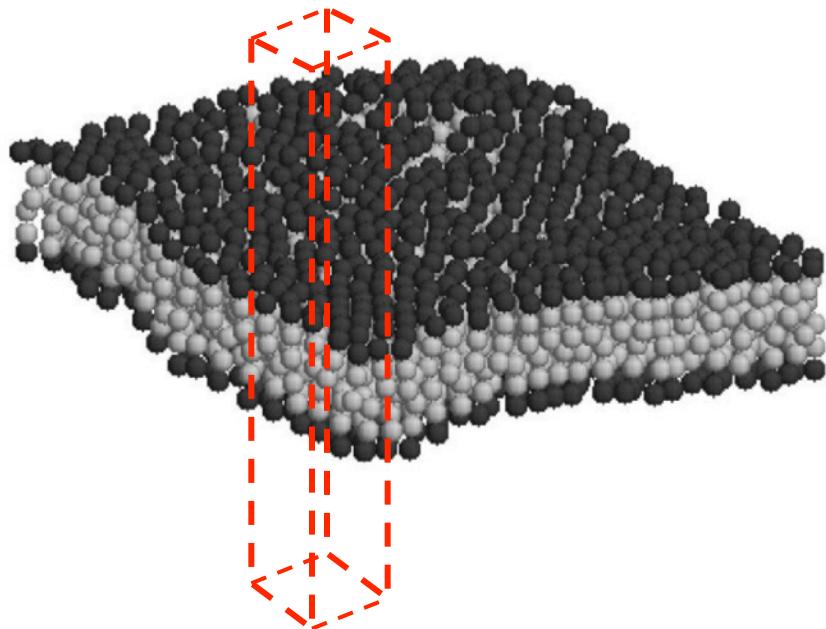
molecules self-assemble spontaneously into planar membranes

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(ensuring $\gamma=0$) are needed
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top view

side view



Properties of a mathematical membrane

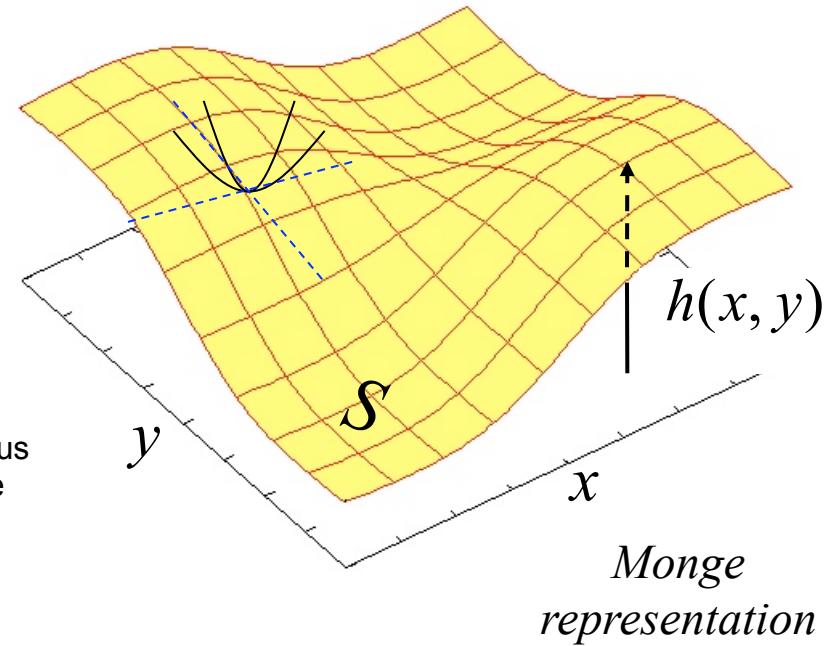
Fluctuating sheet of zero thickness

Described by Helfrich free energy:

$$F = \int_S dA \left[\gamma + \frac{\kappa}{2} \underbrace{(c_1 + c_2)}_{\text{mean curvature}} - 2c_0 \right]^2$$

spontaneous curvature

$$\text{mean curvature } H = \frac{c_1 + c_2}{2}$$



For our membranes $c_0=0$ and

$$F = \int_S dA (\gamma + 2\kappa H^2) \longrightarrow \frac{1}{2} \iint dx dy \left[\gamma |\nabla_{\perp} h|^2 + \kappa (\nabla^2 h)^2 \right]$$

Properties of a mathematical membrane

Fluctuating sheet of zero thickness

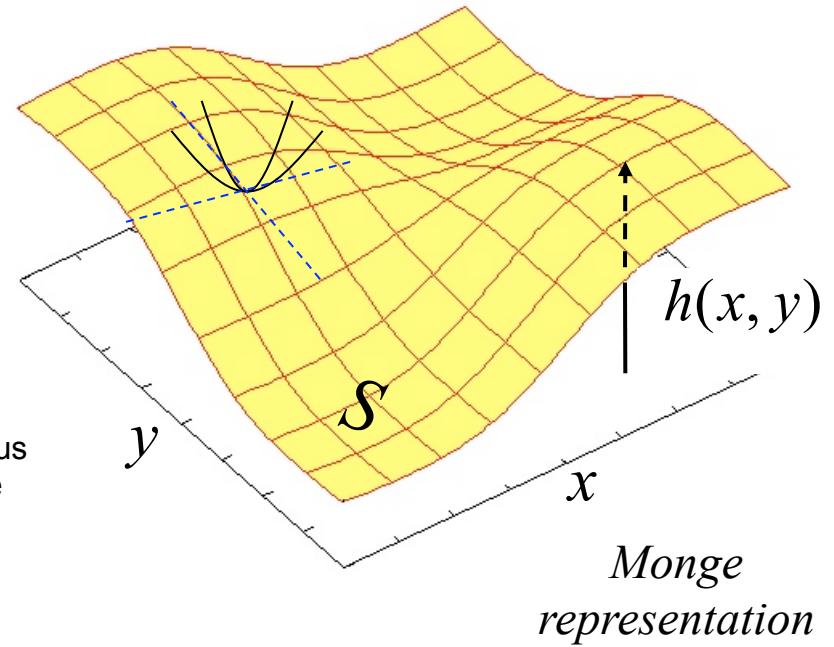
Described by Helfrich free energy:

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surface tension

mean curvature $H = \frac{c_1 + c_2}{2}$

spontaneous curvature



For our membranes $c_0=0$ and

$$F = \int_S dA (\gamma + 2\kappa H^2) \longrightarrow \frac{1}{2} \iint dx dy \left[\gamma |\nabla_\perp h|^2 + \kappa (\nabla^2 h)^2 \right]$$

Properties of a mathematical membrane

Fluctuating sheet of zero thickness

Described by Helfrich free energy:

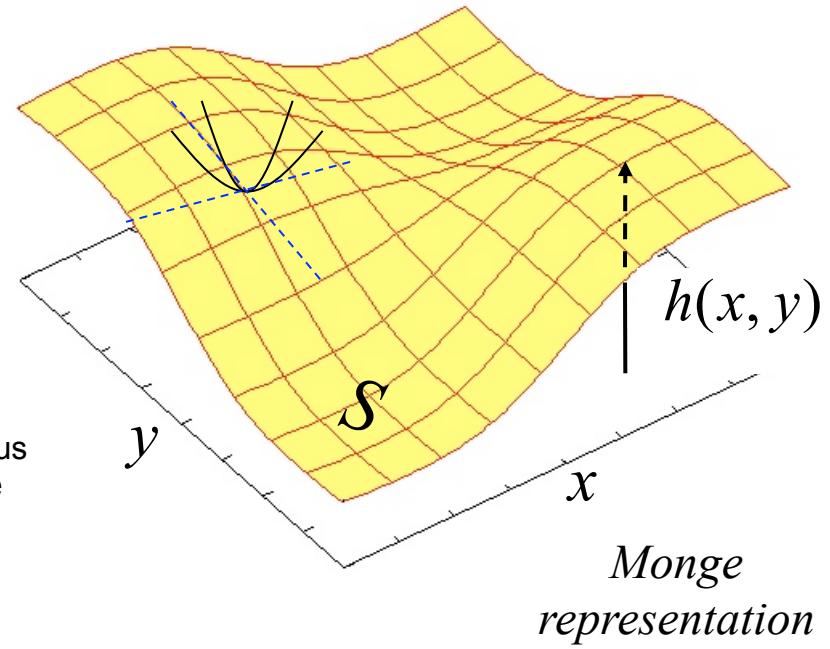
$$F = \int_S dA \left[\gamma + \frac{\kappa}{2} (c_1 + c_2 - 2c_0)^2 \right]$$

surface tension γ

bending constant κ

mean curvature $H = \frac{c_1 + c_2}{2}$

spontaneous curvature c_0



For our membranes $c_0=0$ and

$$F = \int_S dA (\gamma + 2\kappa H^2) \longrightarrow \frac{1}{2} \iint dx dy \left[\gamma |\nabla_{\perp} h|^2 + \kappa (\nabla^2 h)^2 \right]$$

How can γ and κ be computed from a MD simulation?

Taking Fourier transform:

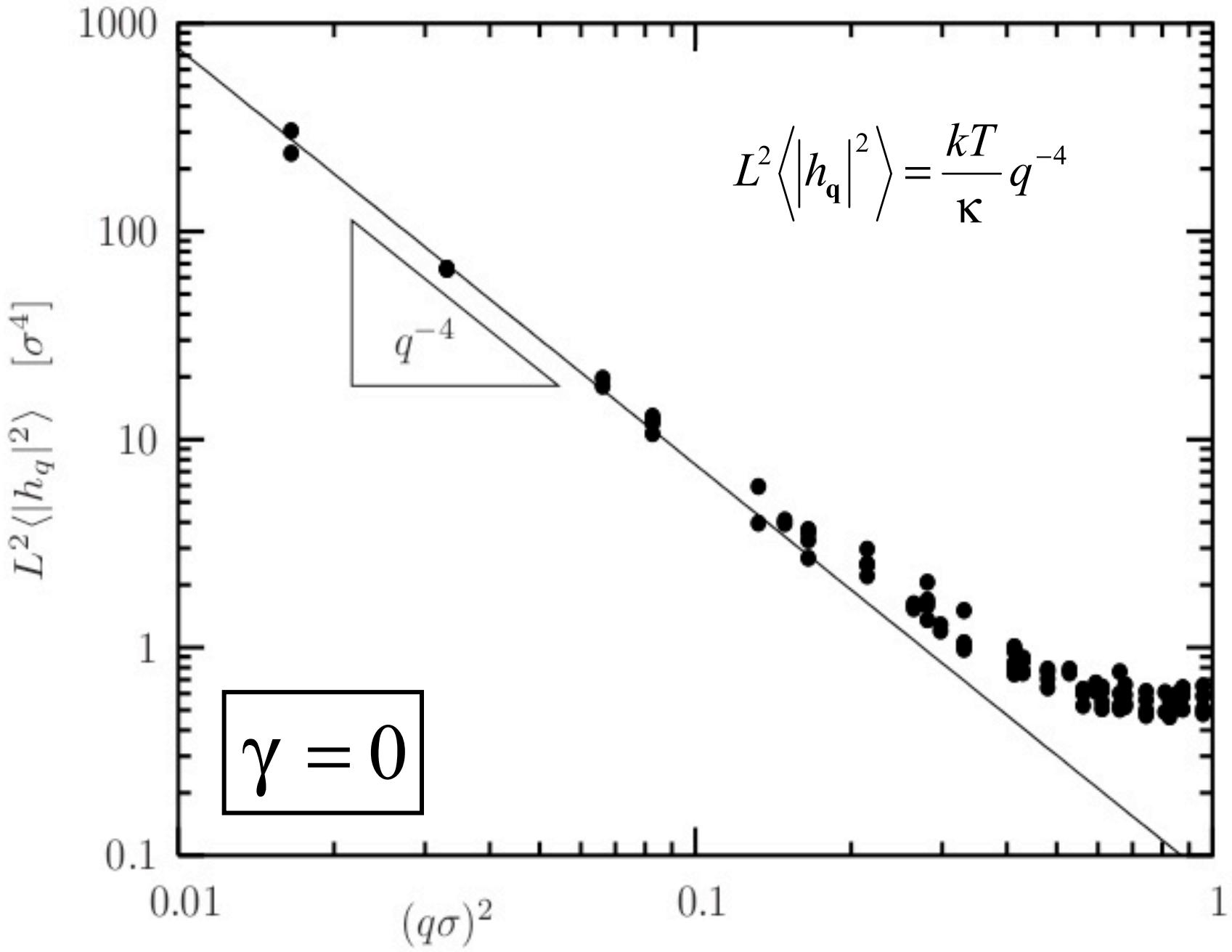
$$F = \frac{L^2}{2} \sum_{\mathbf{q}} \left(\gamma |\mathbf{q}|^2 + \kappa |\mathbf{q}|^4 \right) |h_{\mathbf{q}}|^2$$

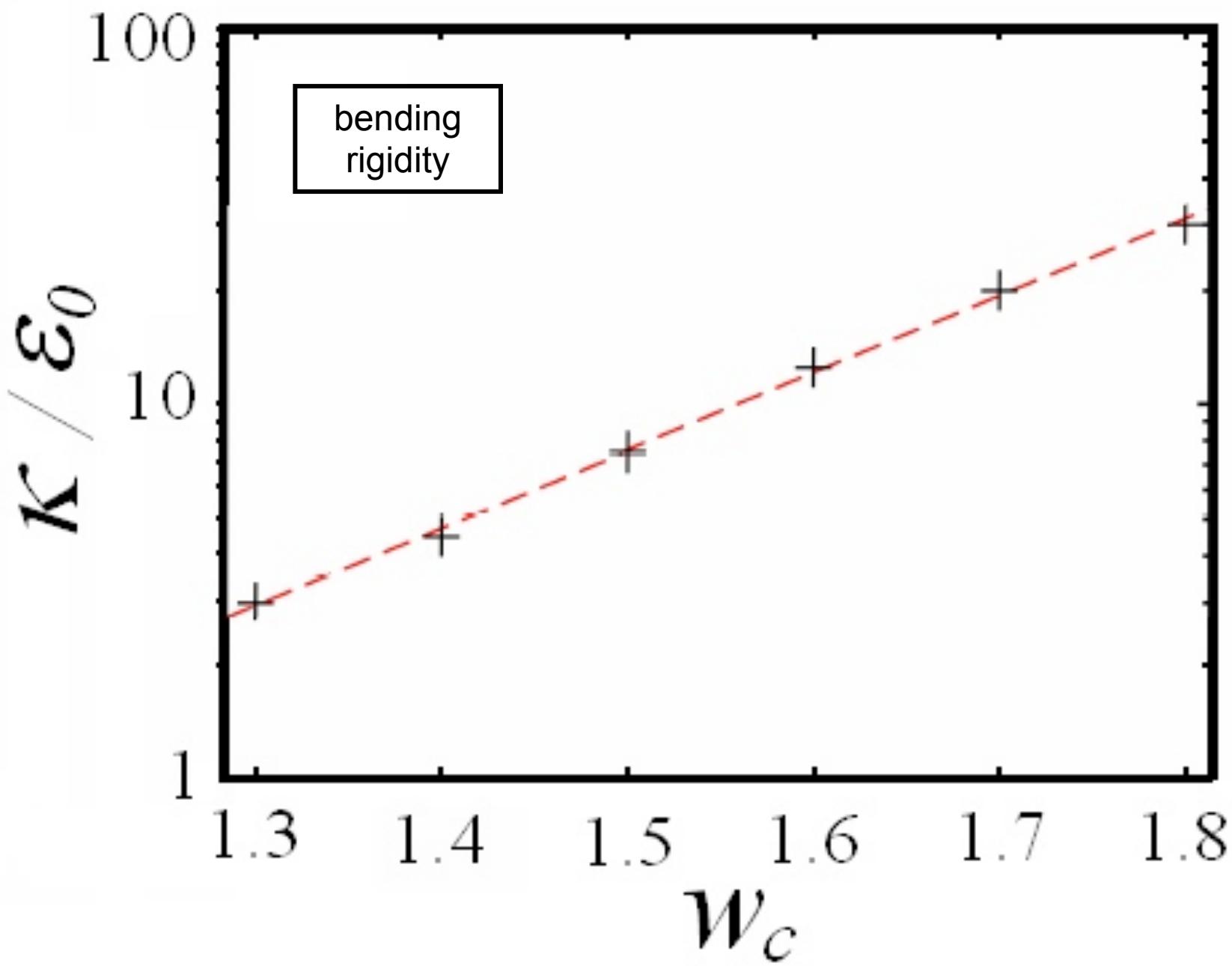
Equipartition theorem:

$$\frac{L^2}{2} \left(\gamma |\mathbf{q}|^2 + \kappa |\mathbf{q}|^4 \right) \langle |h_{\mathbf{q}}|^2 \rangle = \frac{kT}{2}$$



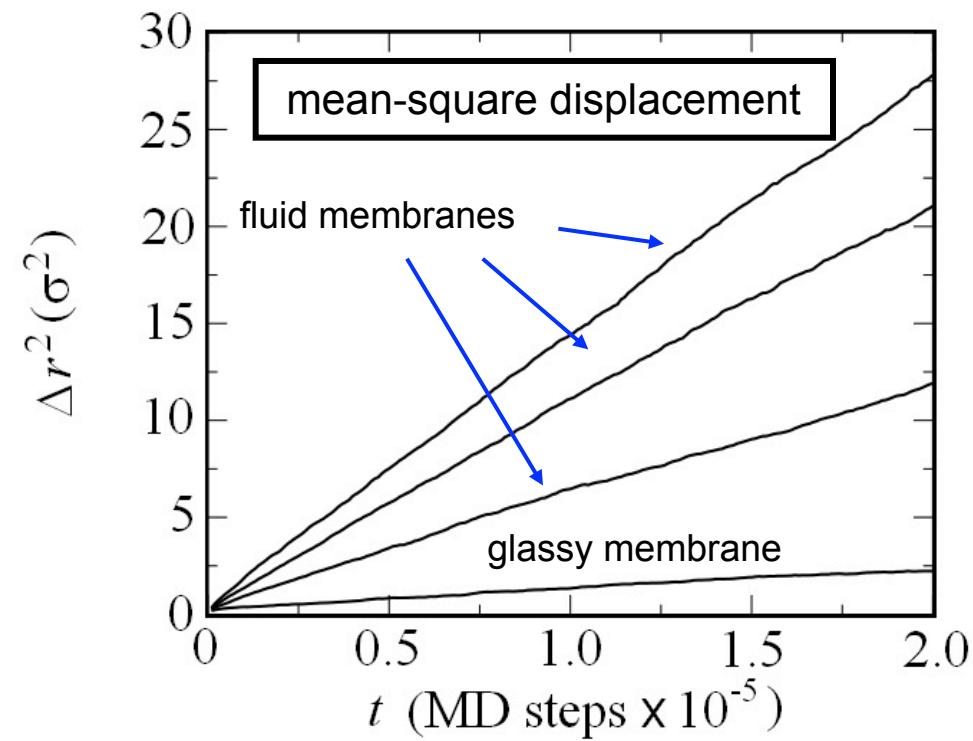
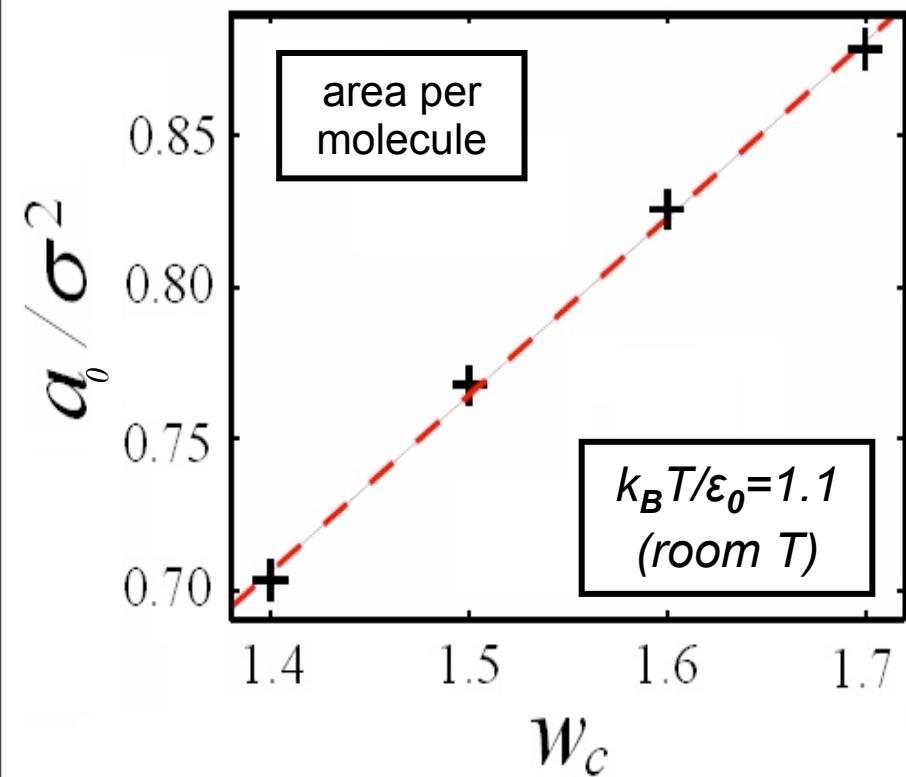
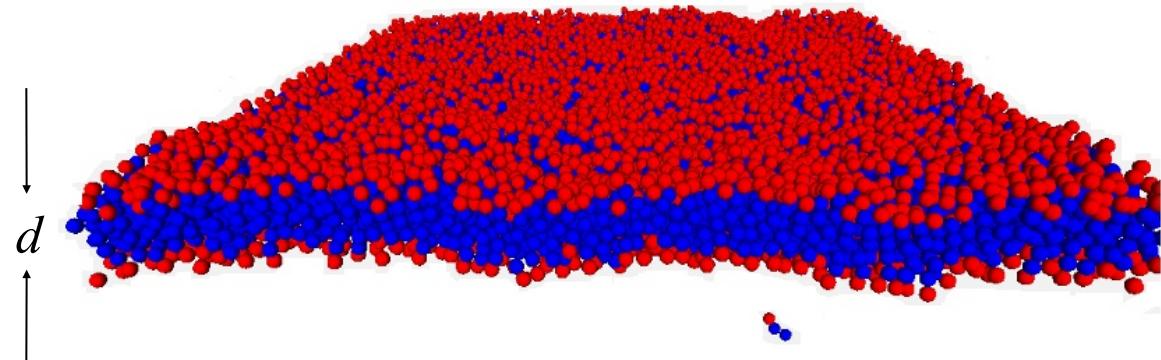
$$L^2 \langle |h_{\mathbf{q}}|^2 \rangle = \frac{kT}{\gamma |\mathbf{q}|^2 + \kappa |\mathbf{q}|^4}$$





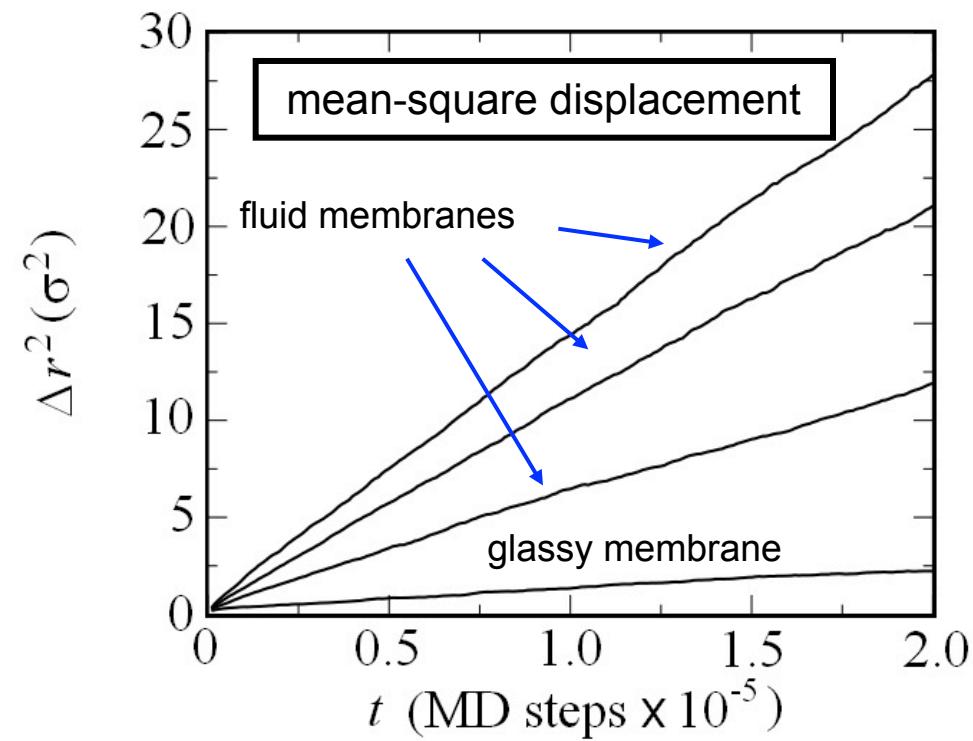
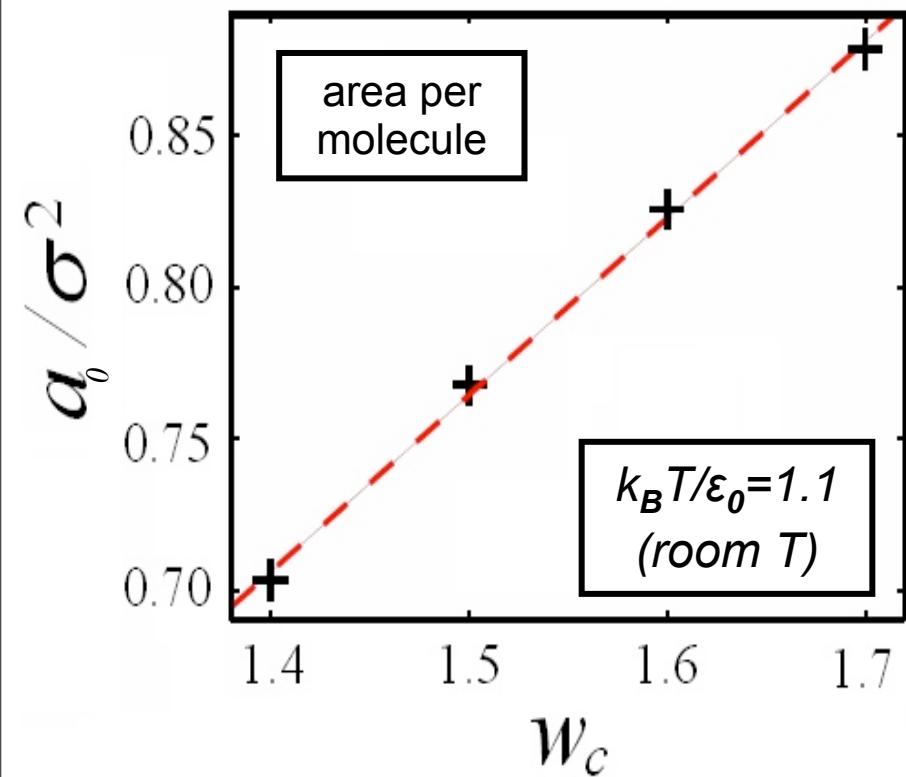
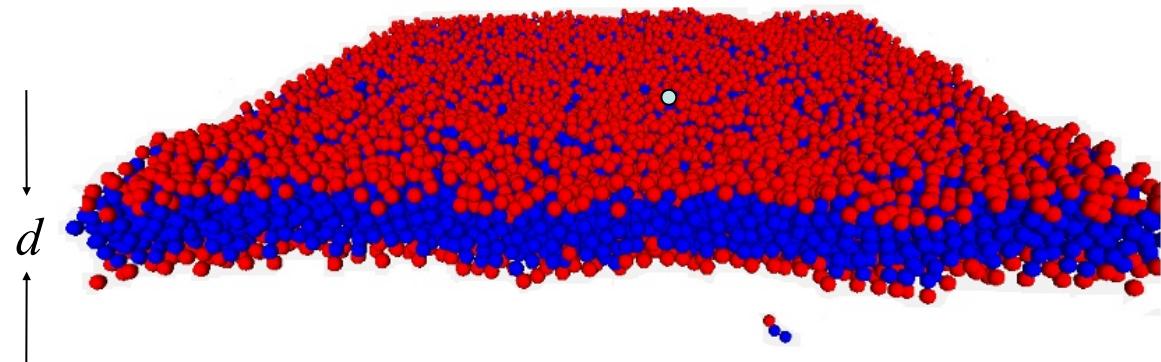
Properties of a physical membrane

- membrane thickness
- two-dimensional density
- molecular diffusion
- lateral compressibility



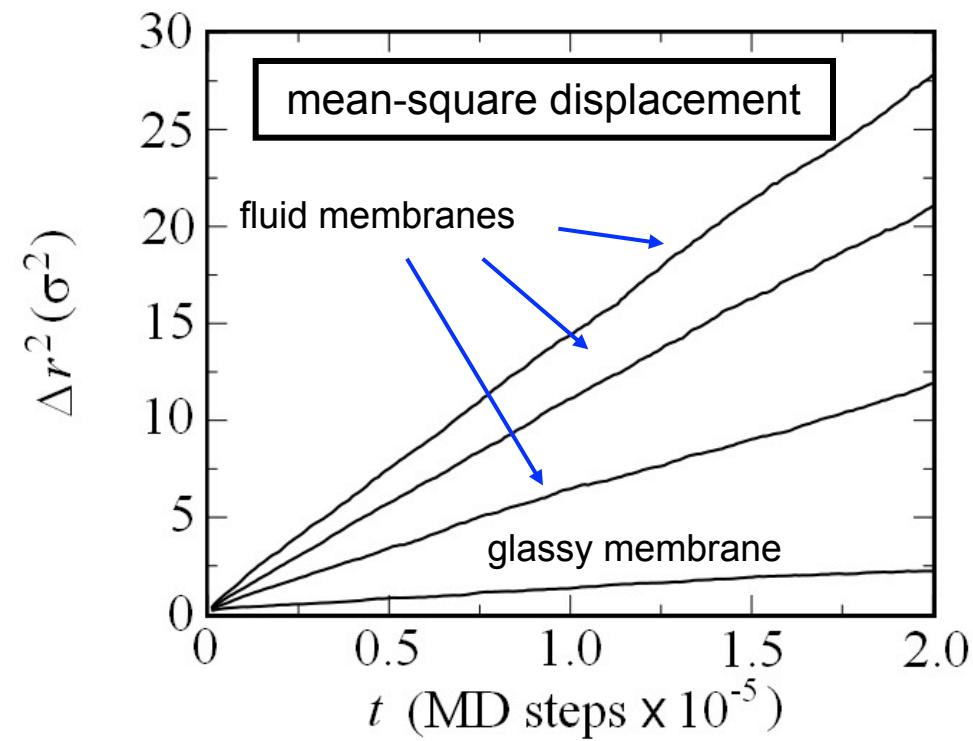
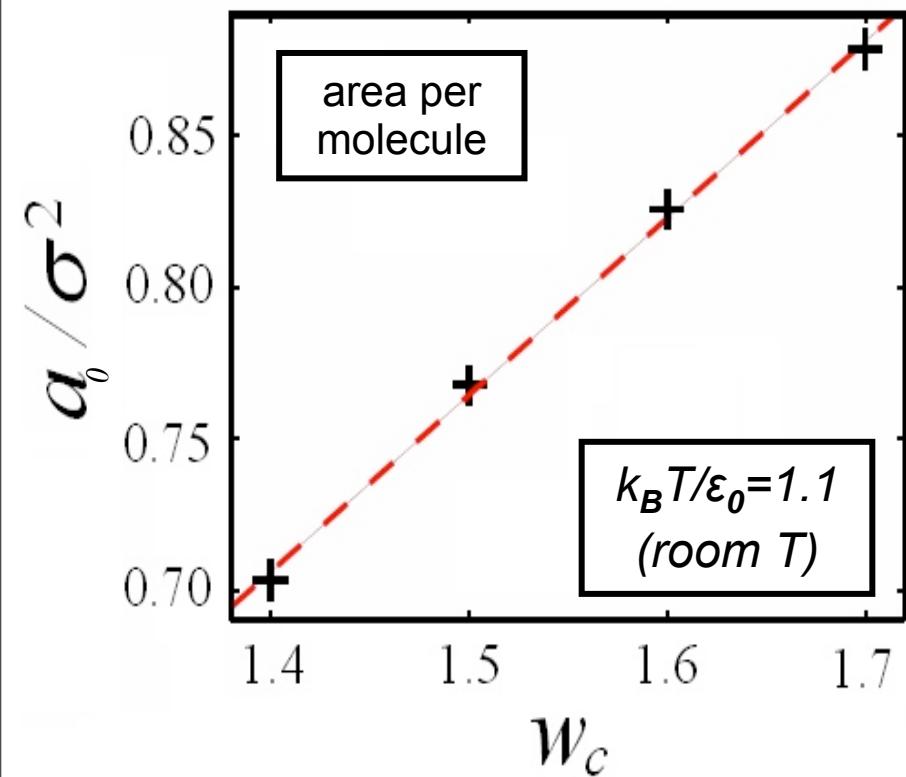
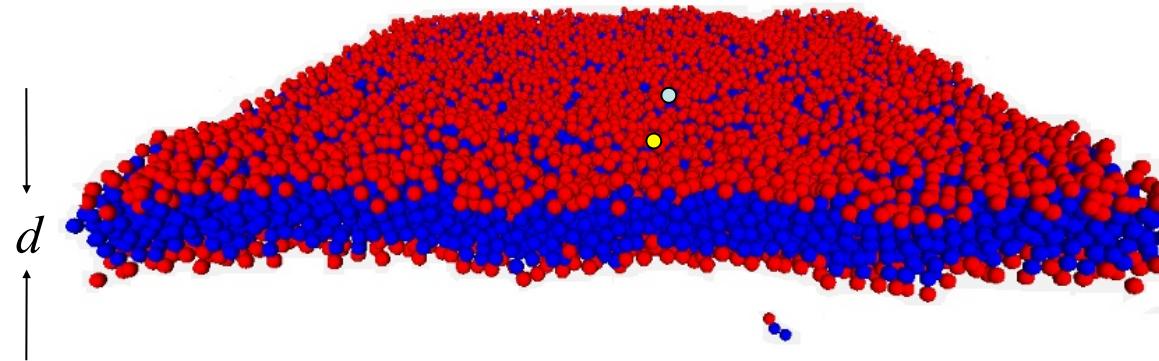
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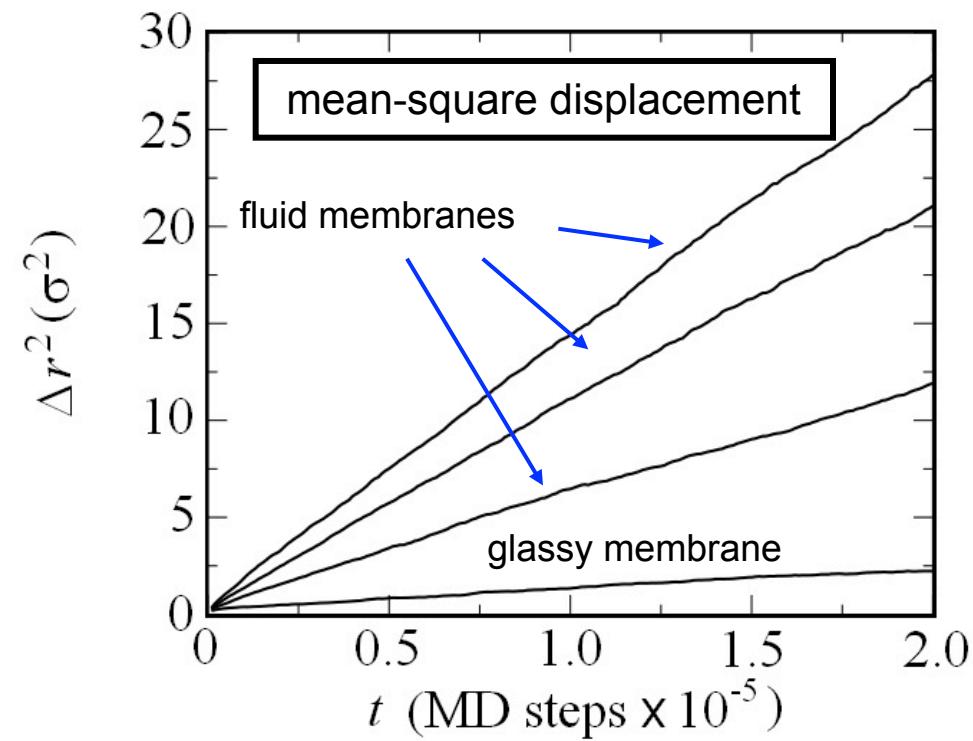
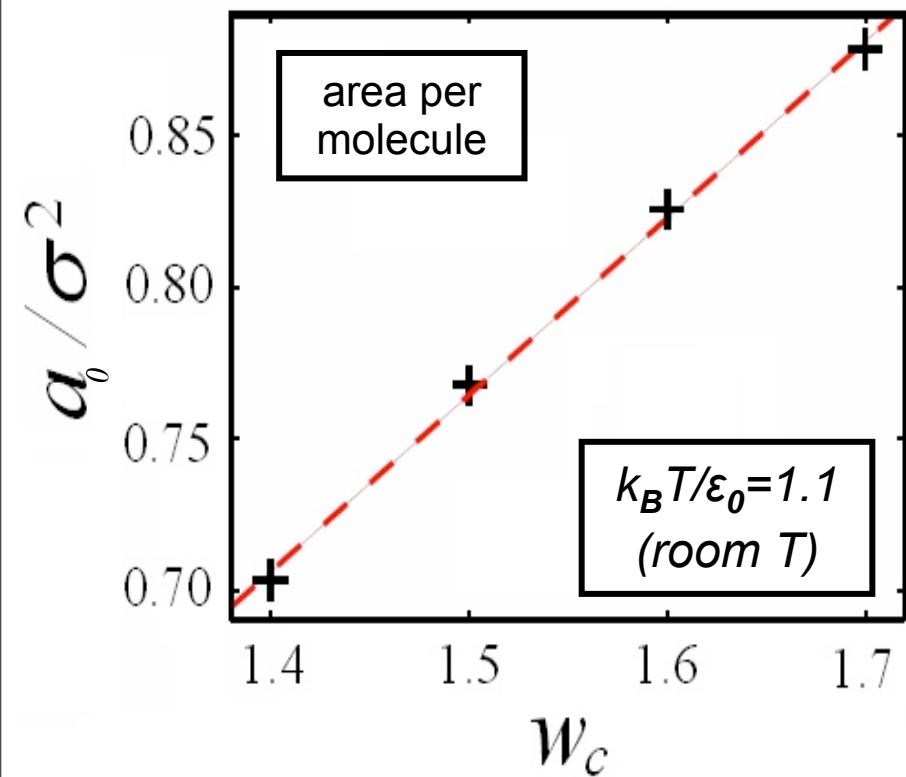
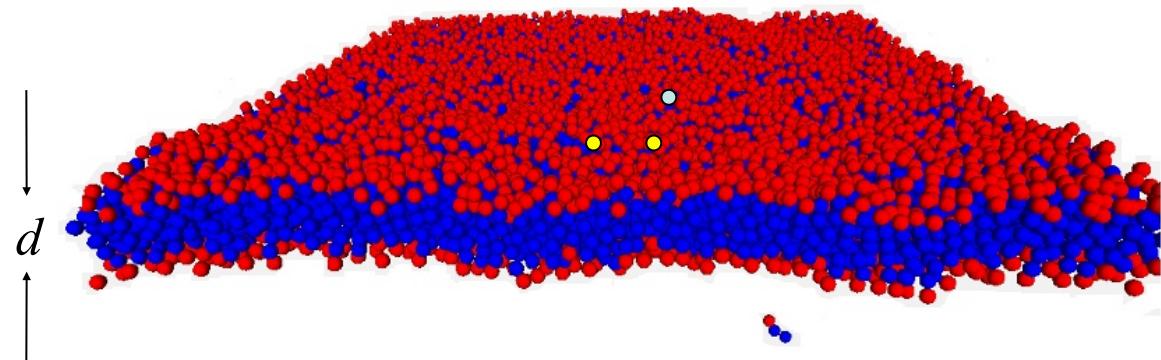
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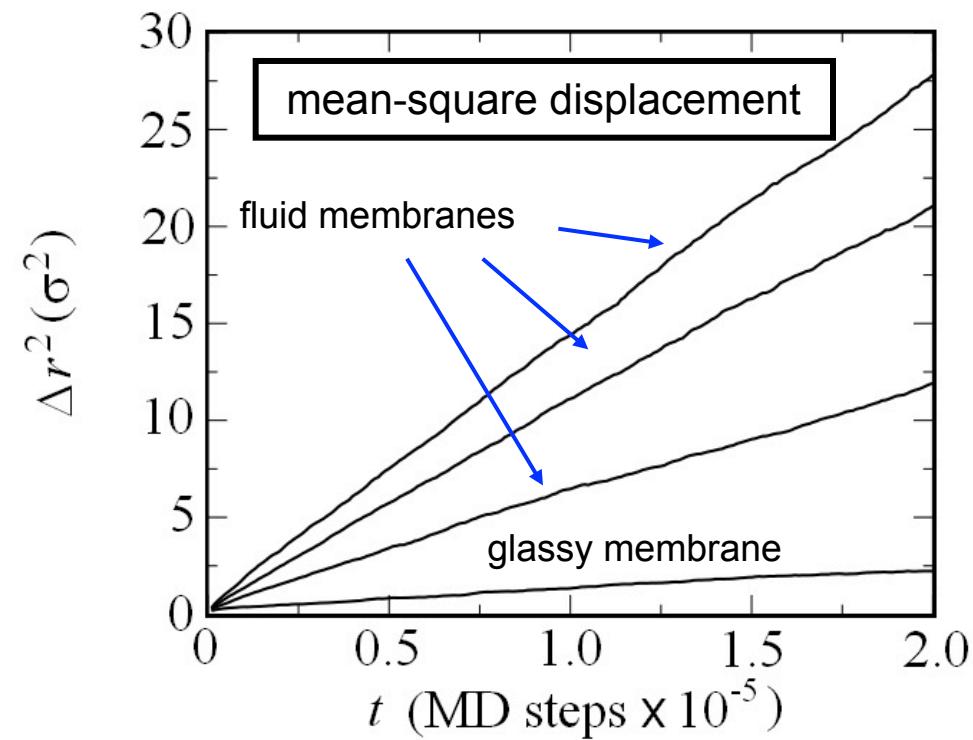
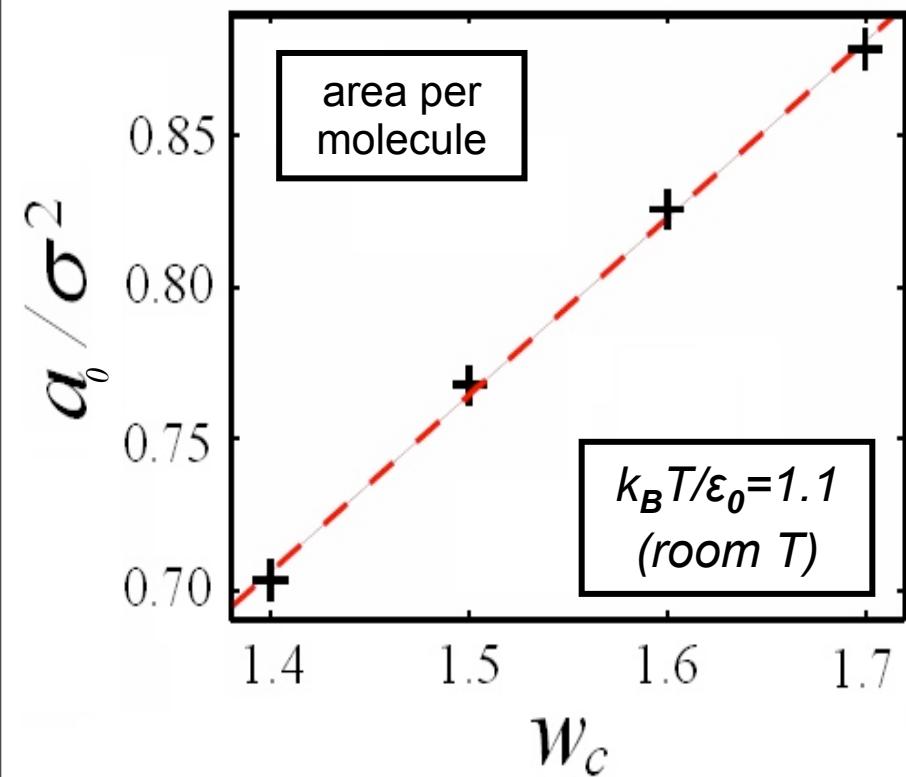
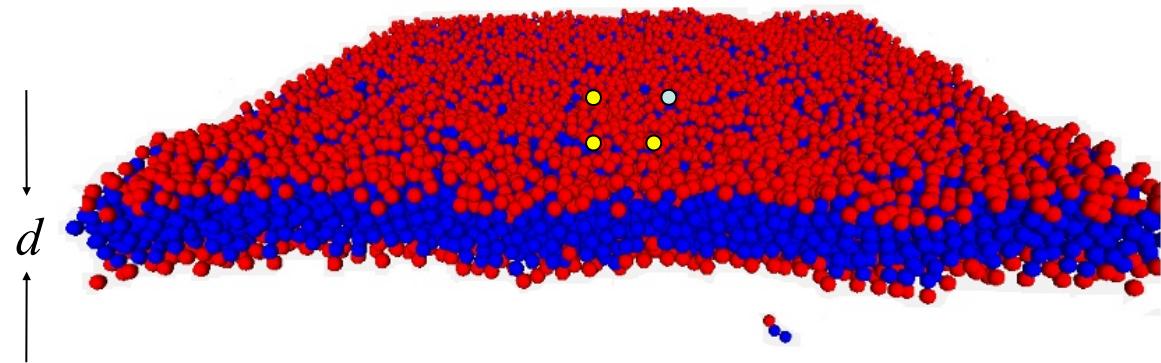
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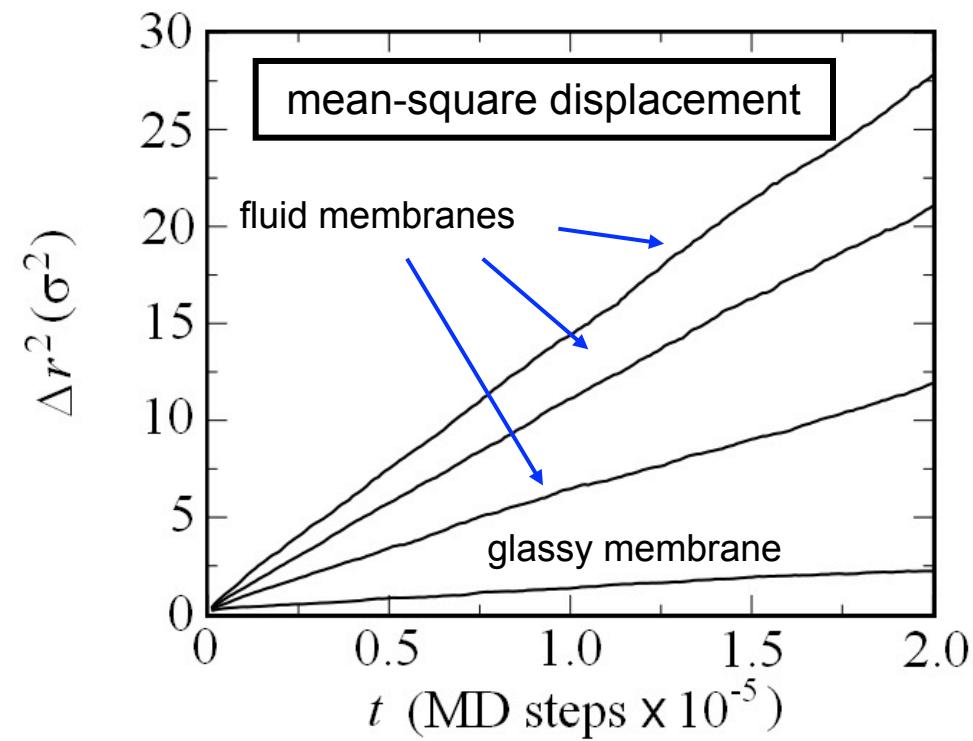
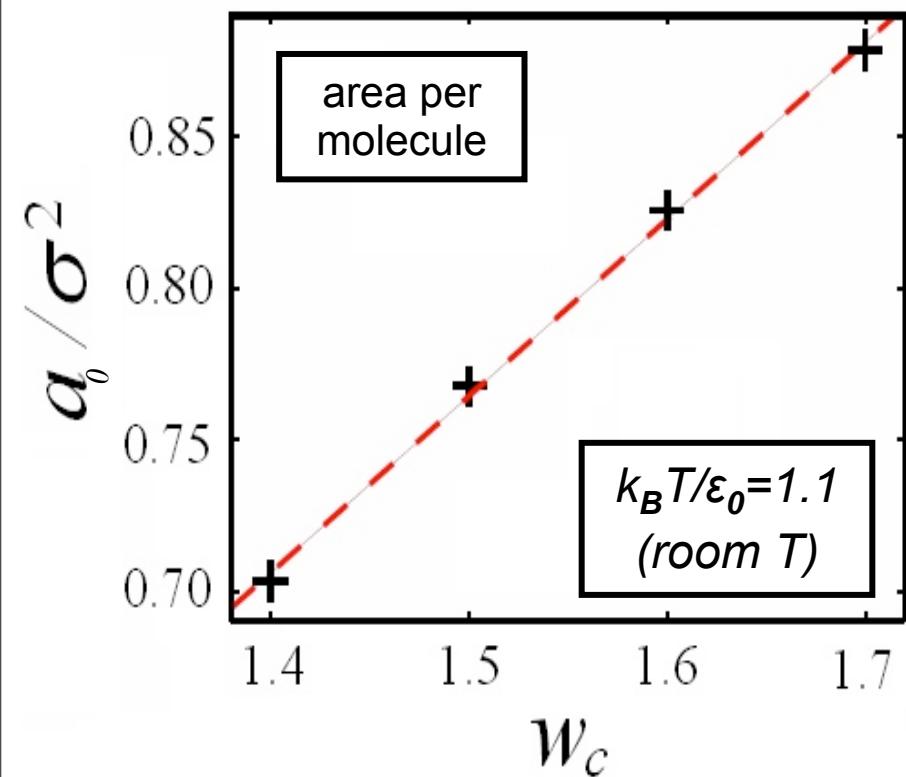
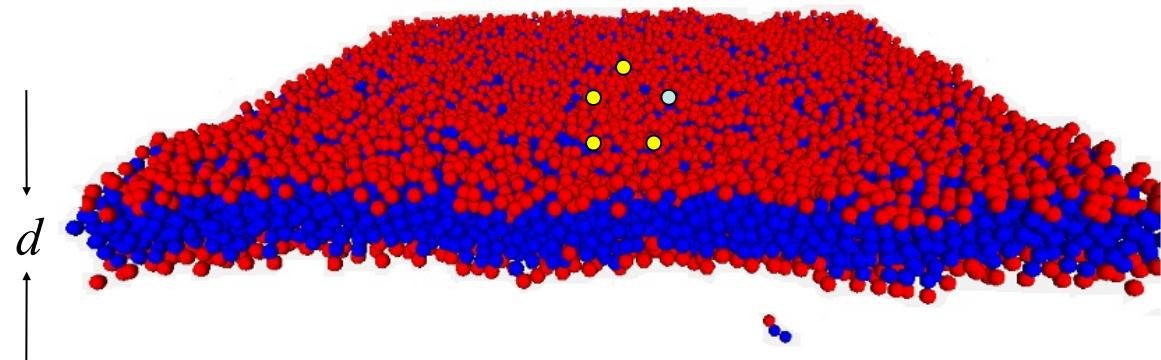
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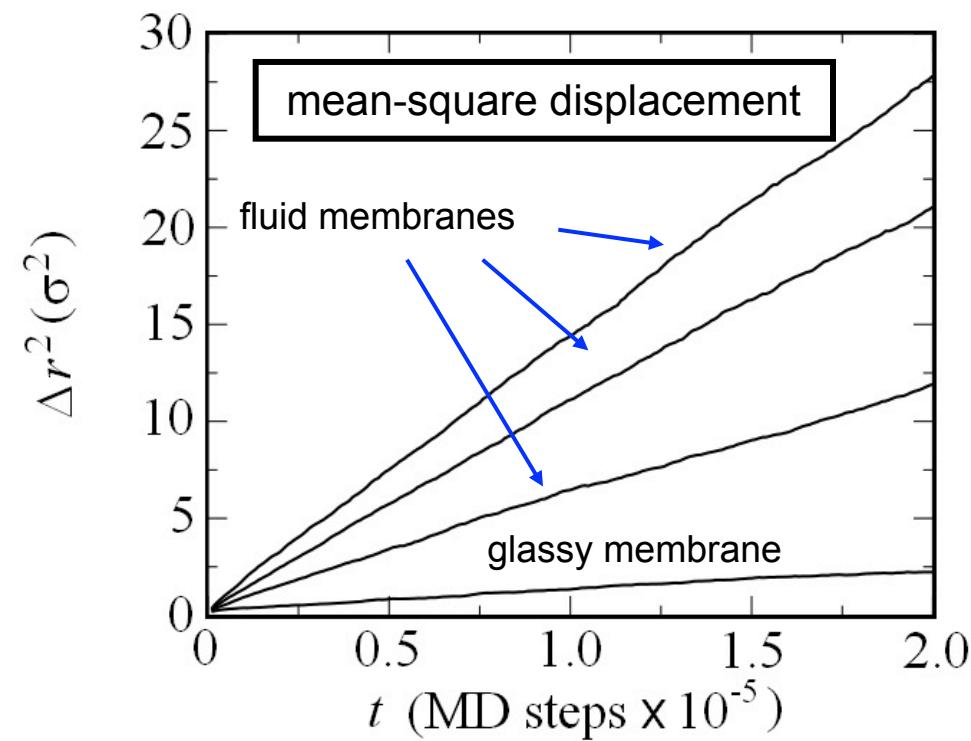
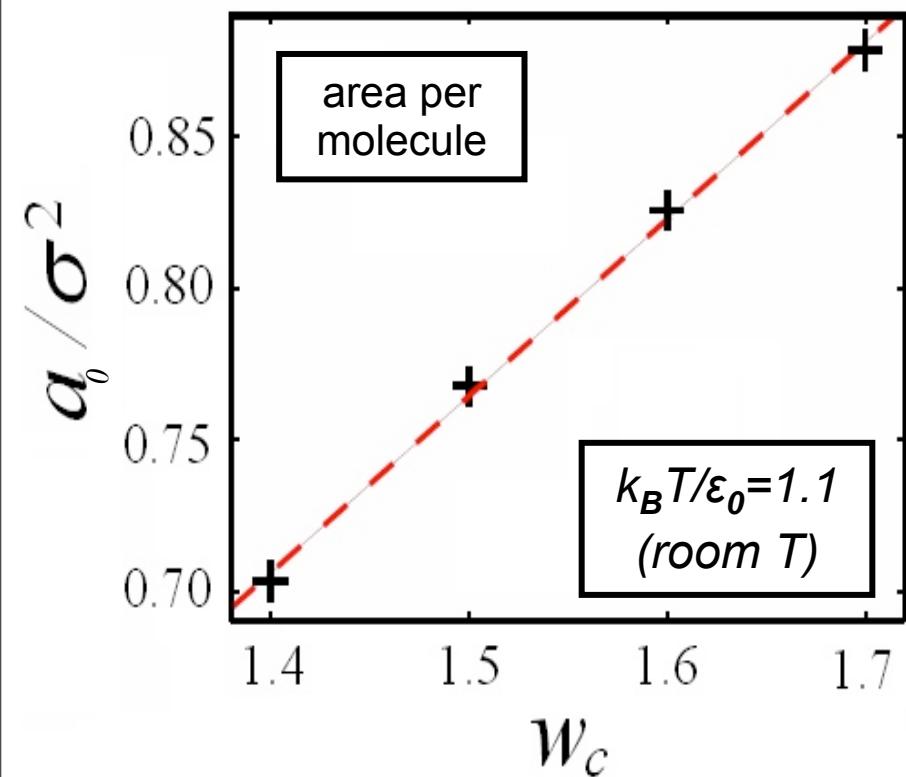
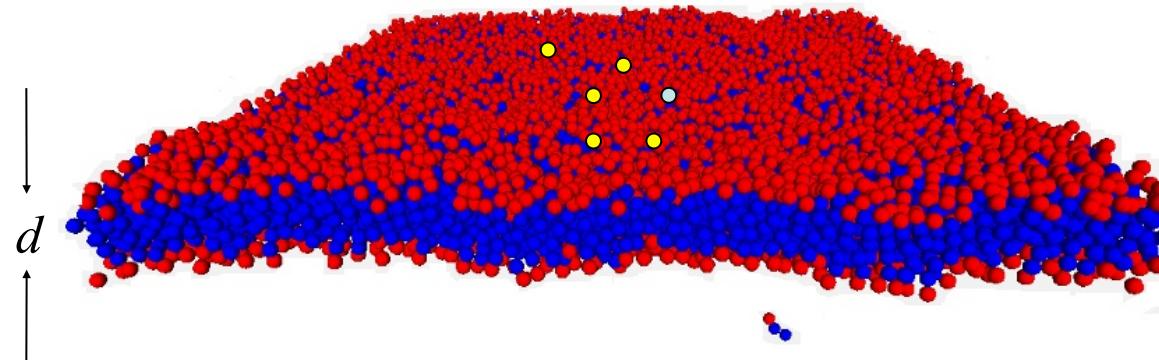
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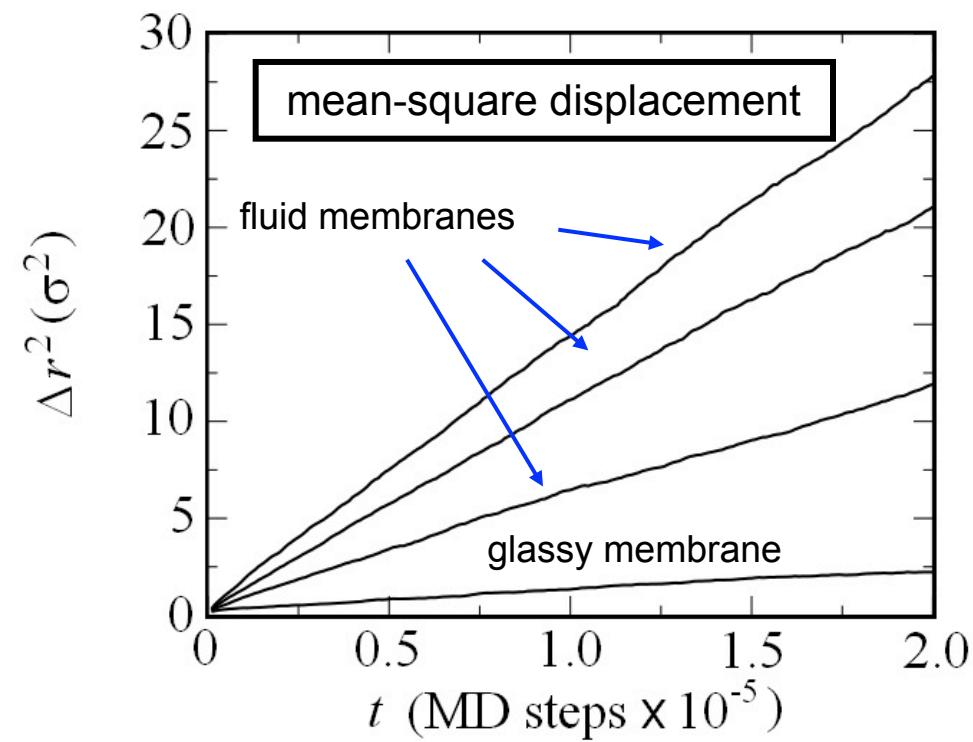
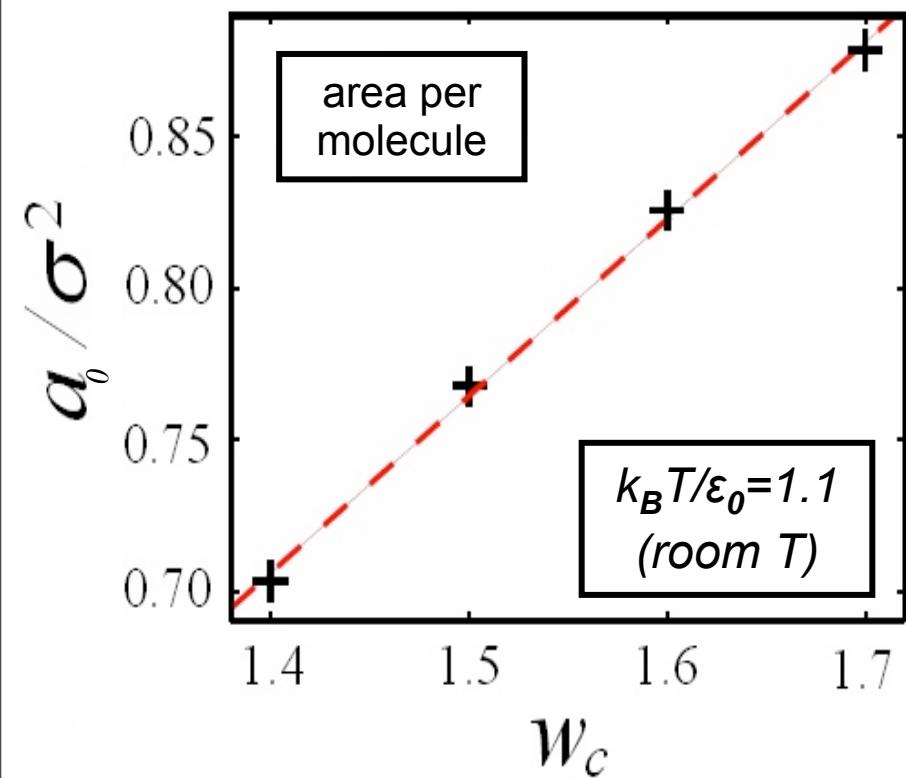
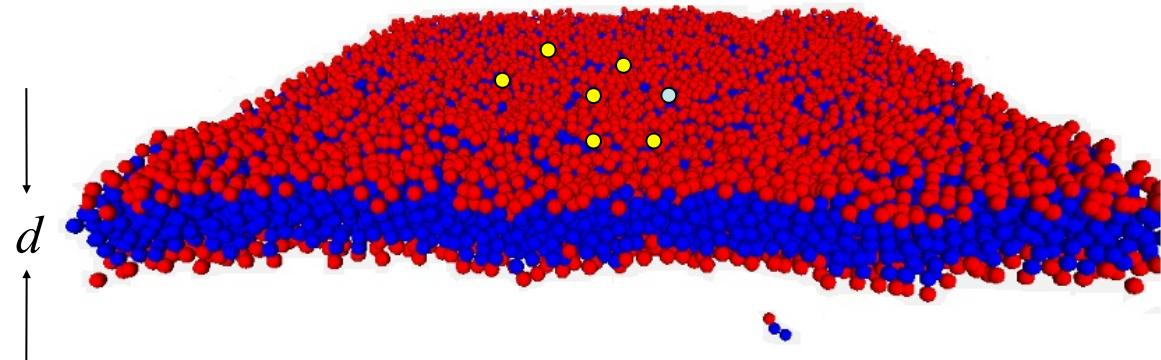
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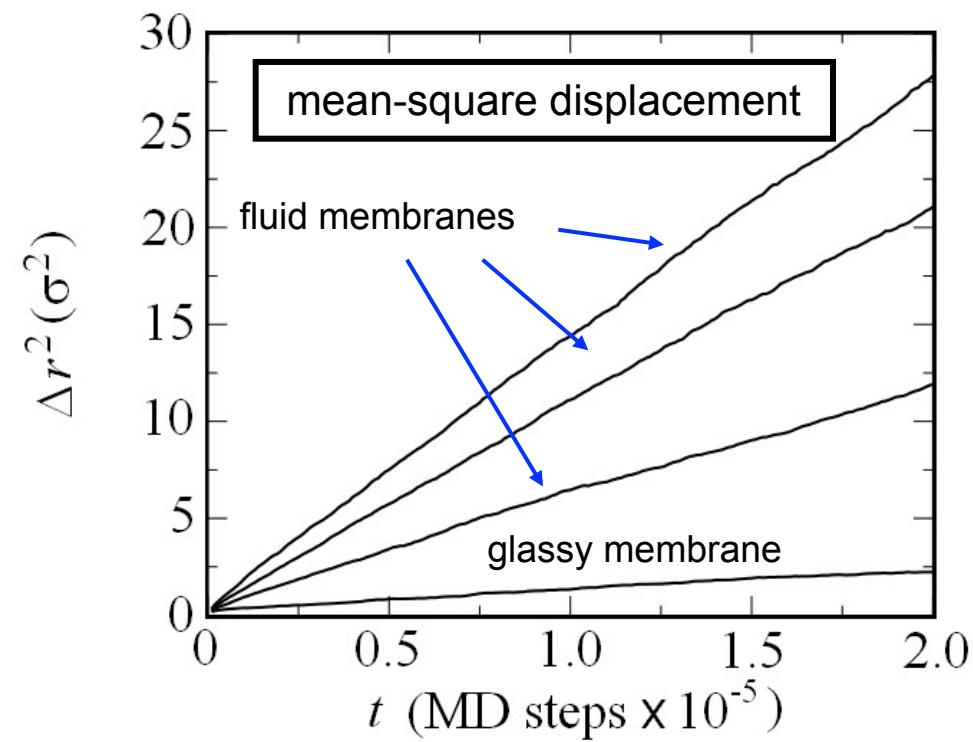
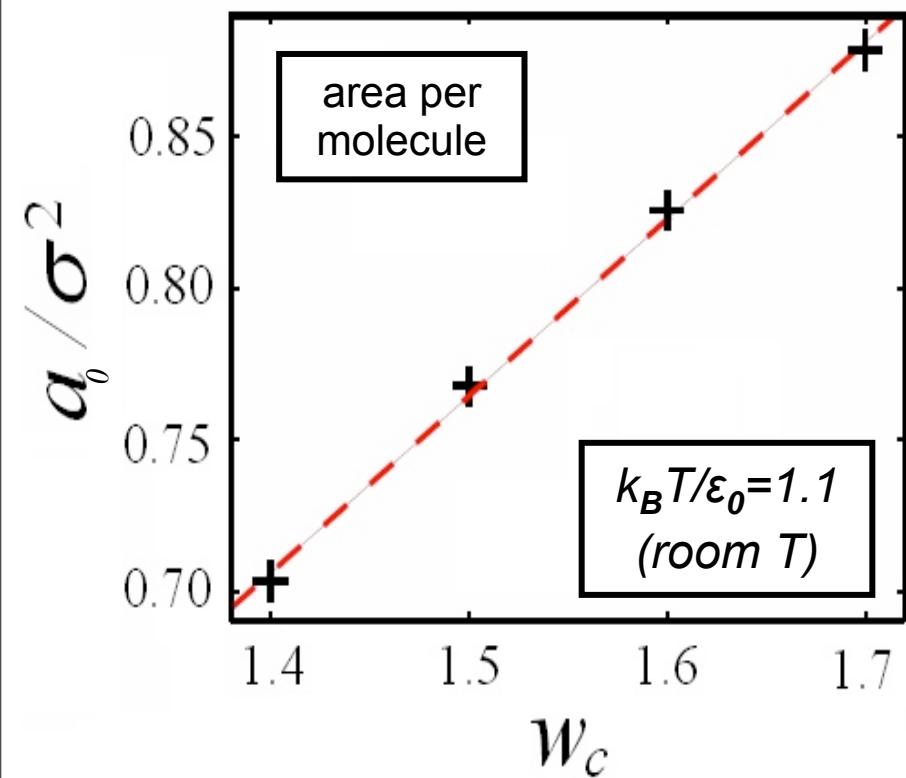
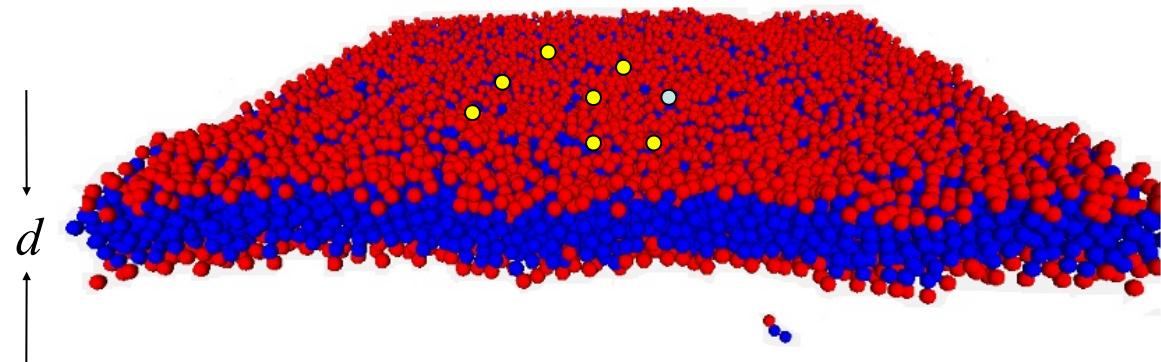
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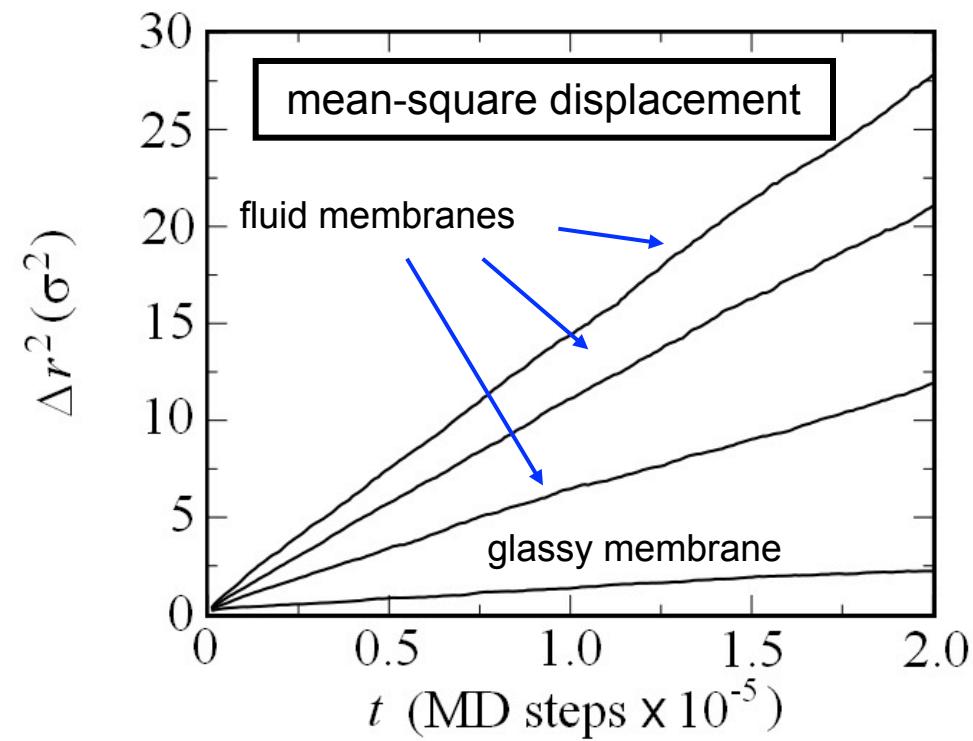
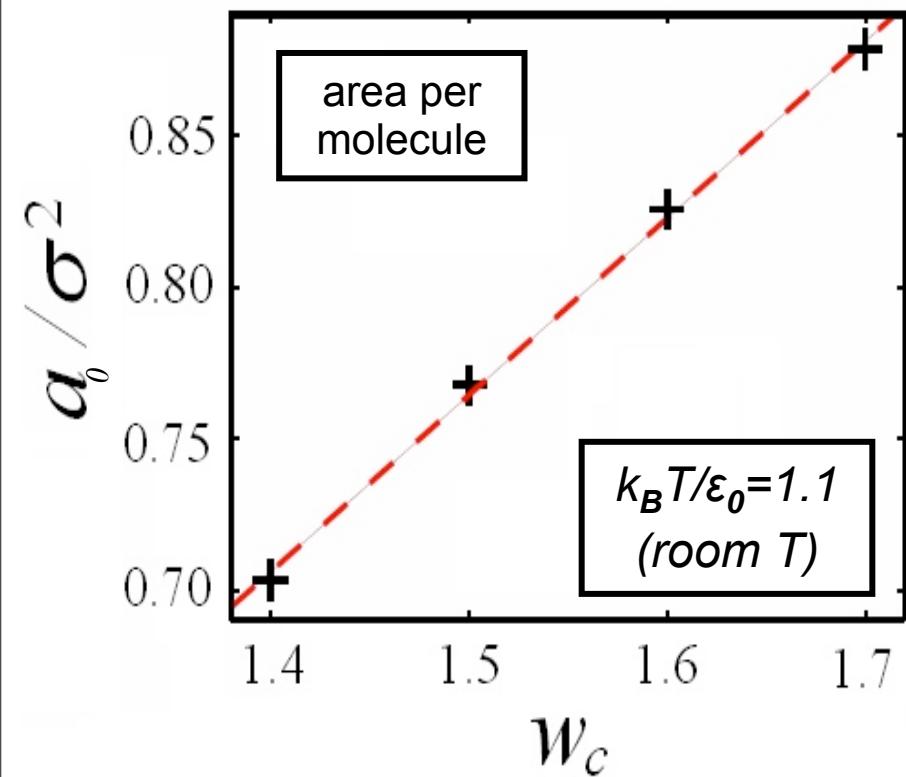
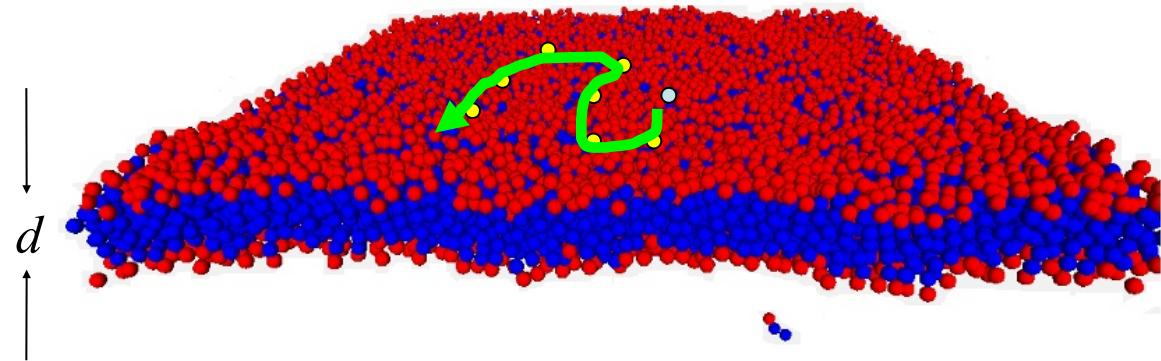
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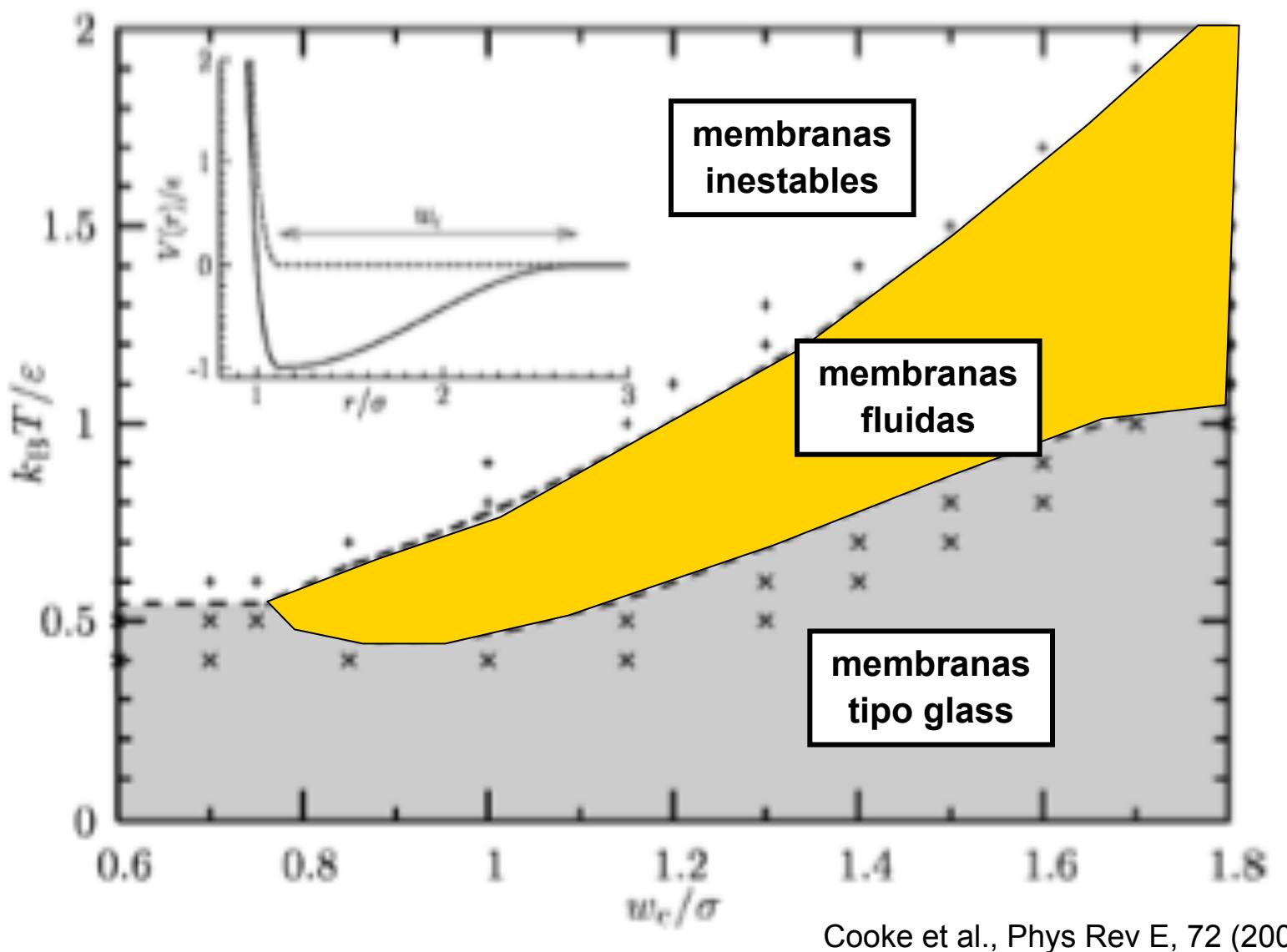


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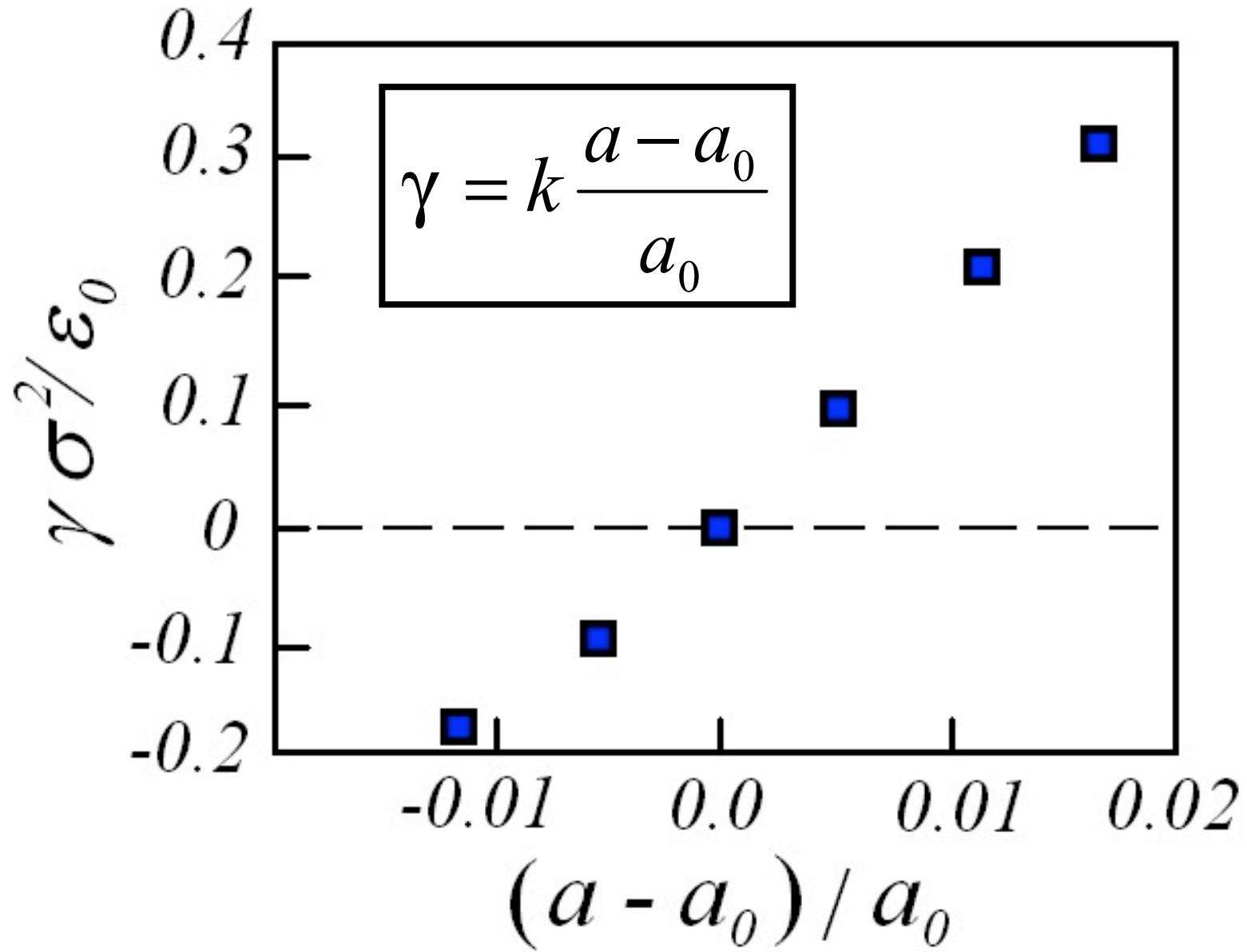


Influencia del parámetro de alcance w_c



Cooke et al., Phys Rev E, 72 (2005)

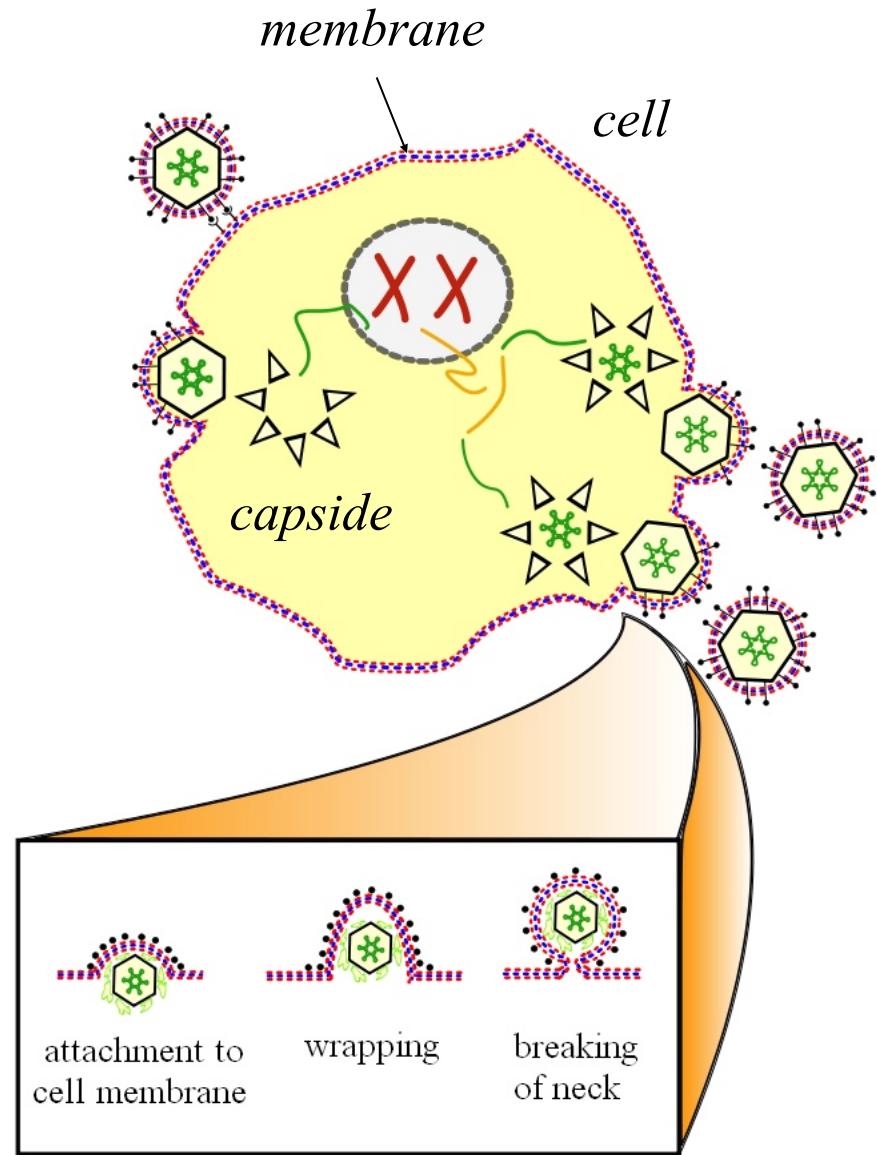
lateral compressibility, k



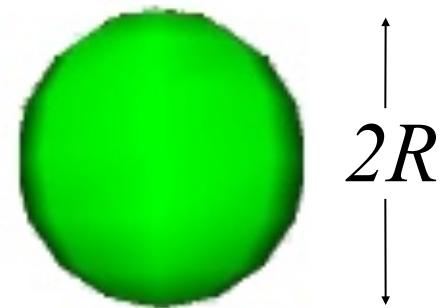
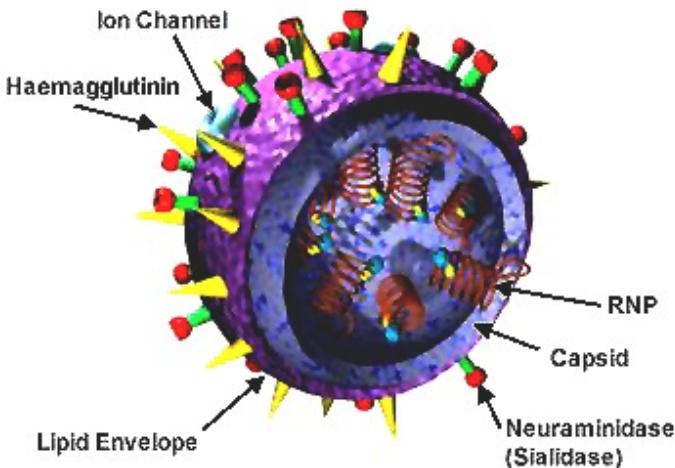
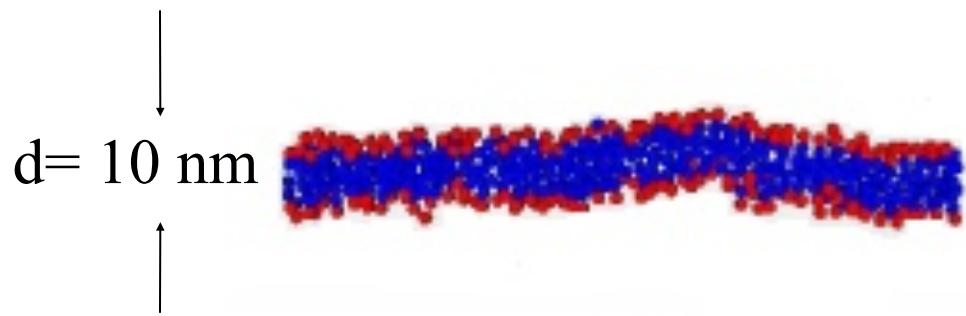
Application: wrapping and budding of viruses

After using the cell machinery to replicate, some viruses leave the cell by acquiring a membrane coating

We aim at understanding the physics of the problem using a very simplified model



enveloped
virus



→ $40 - 600 \text{ nm}$ ←

$$\frac{2R}{d} = 4 - 60.$$

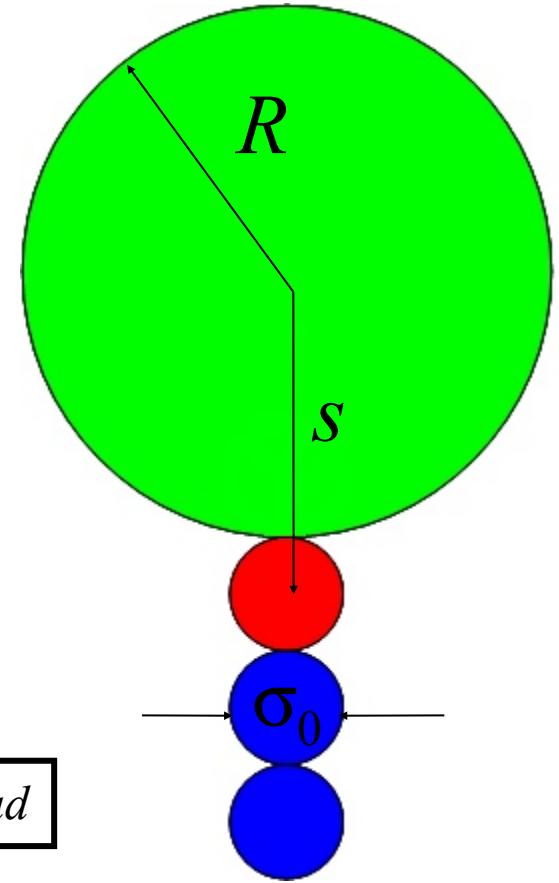
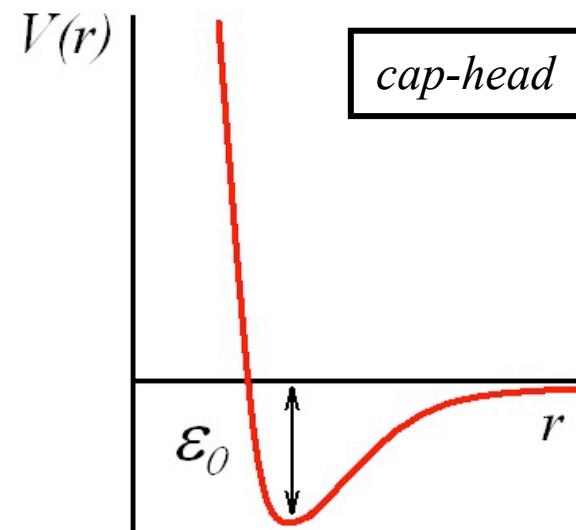
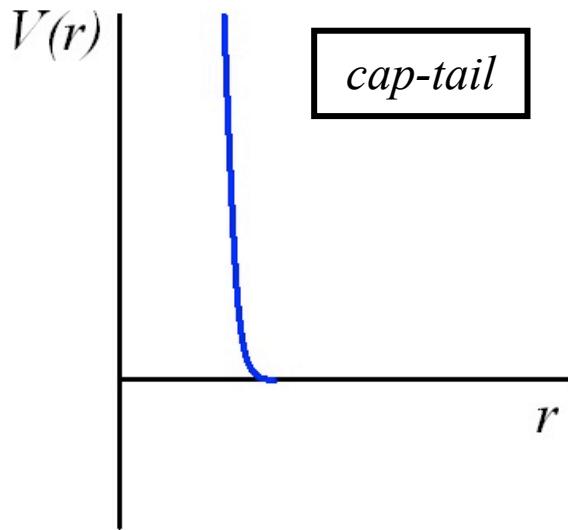
Since $d = 6\sigma_0$, we have:

$$R = 12 - 180\sigma_0$$

Membrane-particle interaction:

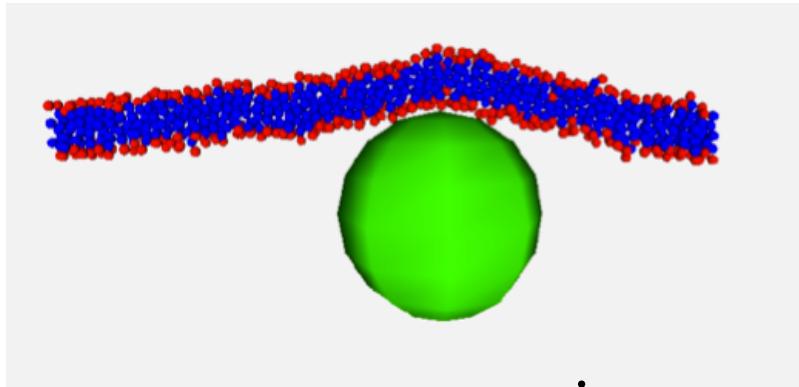
$$V_{cap-tail}(r) = 4\epsilon_0 \begin{cases} \left[\left(\frac{\sigma_0}{r-s} \right)^{12} - \left(\frac{\sigma_0}{r-s} \right)^6 + \frac{1}{4} \right], & r < r_c, \\ 0, & r > r_c. \end{cases}$$

$$V_{cap-head}(r) = 4\epsilon_s \left[\left(\frac{\sigma_0}{r-s} \right)^{12} - \left(\frac{\sigma_0}{r-s} \right)^6 \right]$$

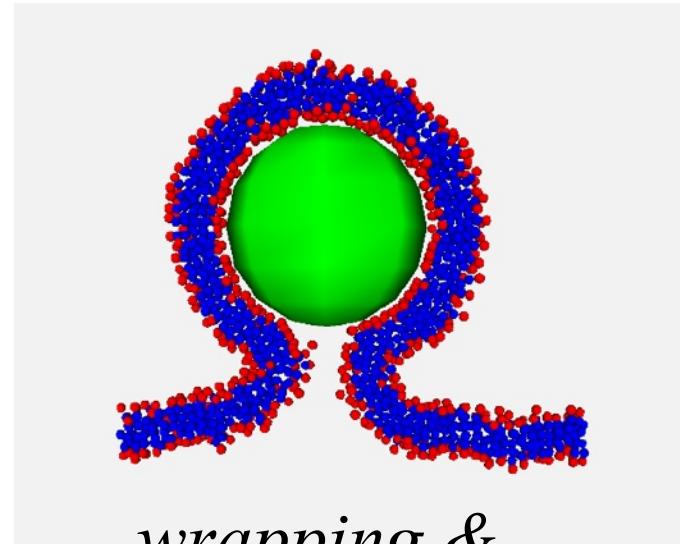


$$s = R + \frac{\sigma_0}{2}$$

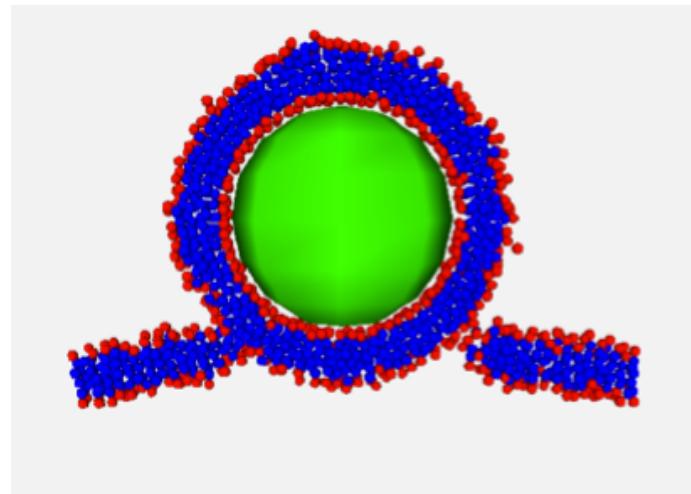
Typical behaviours



non-wrapping



*wrapping &
budding*

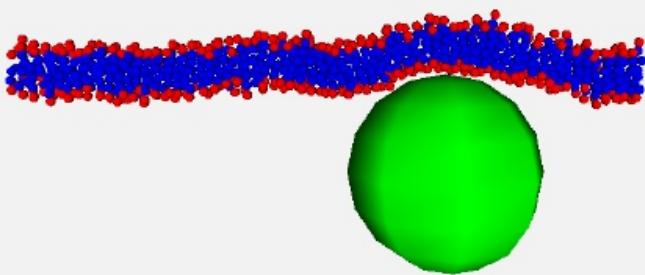


*membrane
breaking*

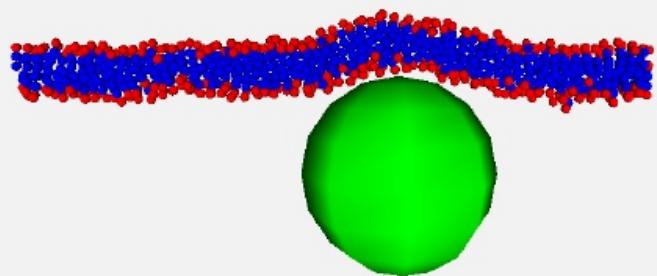
NON - WRAPPING

$\tau_0 = 9.85 \text{ ps}$

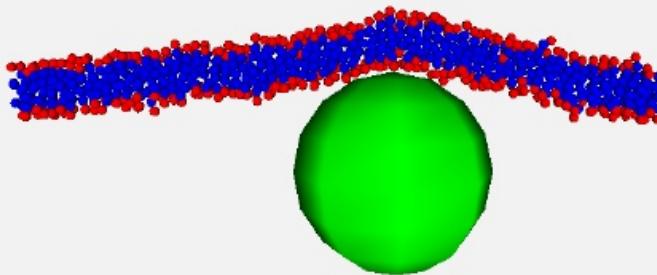
$R = 10 \sigma_0$



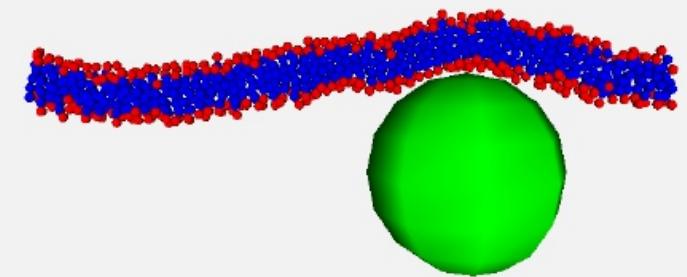
$\tau/\tau_0 = 10^3$



$\tau/\tau_0 = 5 \times 10^3$

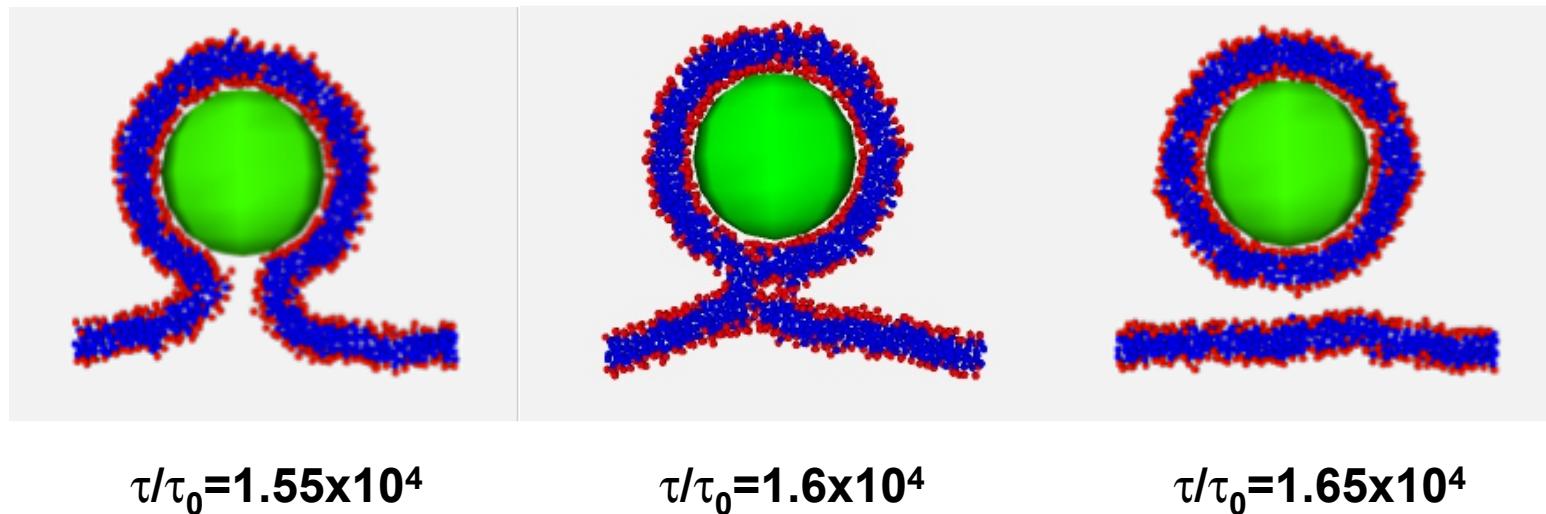
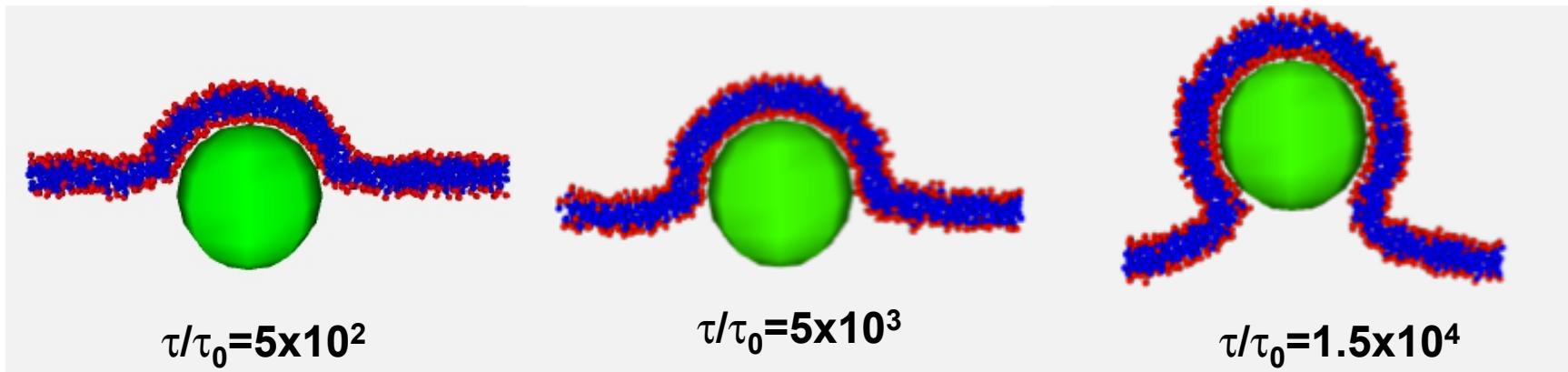


$\tau/\tau_0 = 10^4$

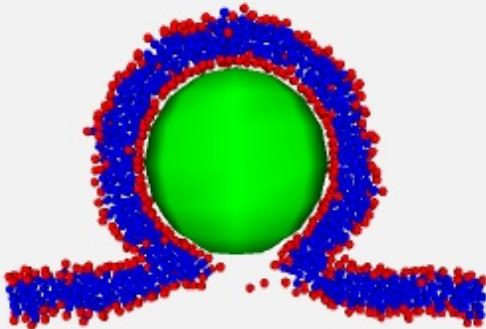


$\tau/\tau_0 = 3 \times 10^4$

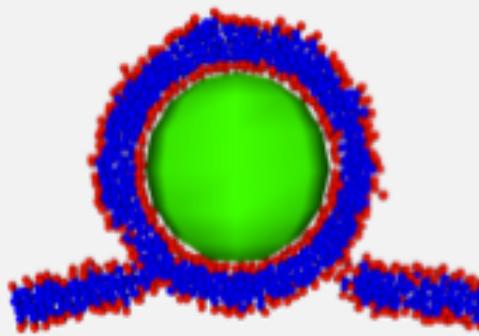
WRAPPING



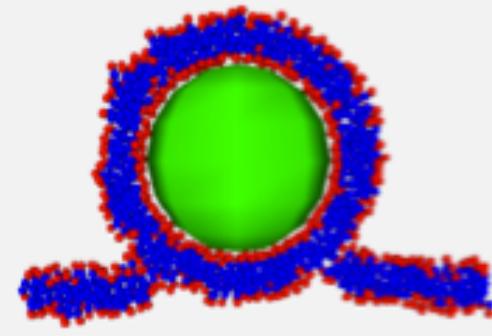
MEMBRANE BREAKING



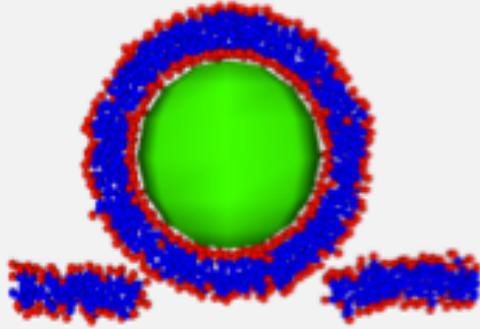
$$\tau/\tau_0 = 5.5 \times 10^3$$



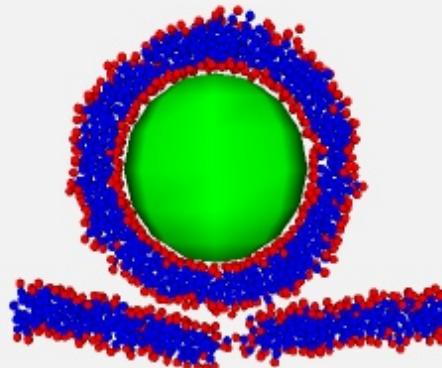
$$\tau/\tau_0 = 6 \times 10^3$$



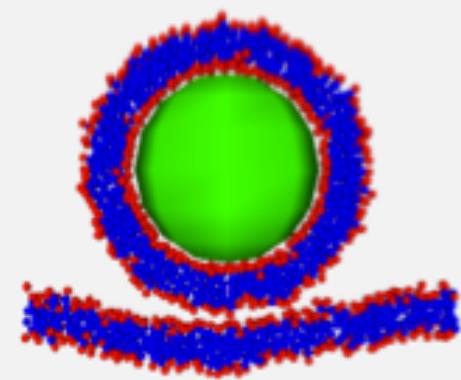
$$\tau/\tau_0 = 7 \times 10^3$$



$$\tau/\tau_0 = 7.5 \times 10^3$$



$$\tau/\tau_0 = 9.5 \times 10^3$$



$$\tau/\tau_0 = 10^4$$

