

# Mesoscopic and Monte Carlo Simulations



Marcus Müller

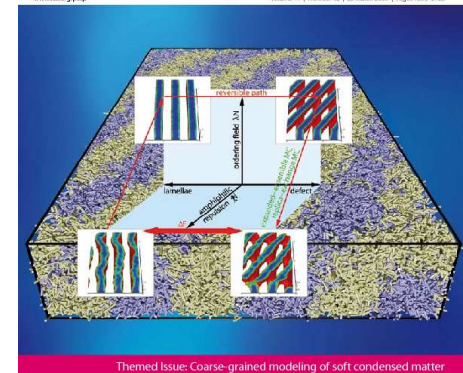
- recap of basic concepts (Mikhail Tamm, Alexey Shaytan): thermodynamics, statistical mechanics, phase transitions
- Monte-Carlo simulations: detailed balance, force-bias Monte-Carlo, phase coexistence
- free-energy of soft-condensed matter: thermodynamic integration in a spatially varying field
- speeding-up intrinsically slow processes: field-theoretic umbrella sampling and HMM
- finding barriers via the on-the-fly string method

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Themed Issue: Coarse-grained modeling of soft condensed matter

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COVER ARTICLE  
Müller et al.  
Computing free energies of interfaces in self-assembling systems

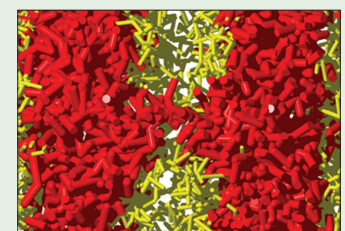
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# basics: thermodynamics

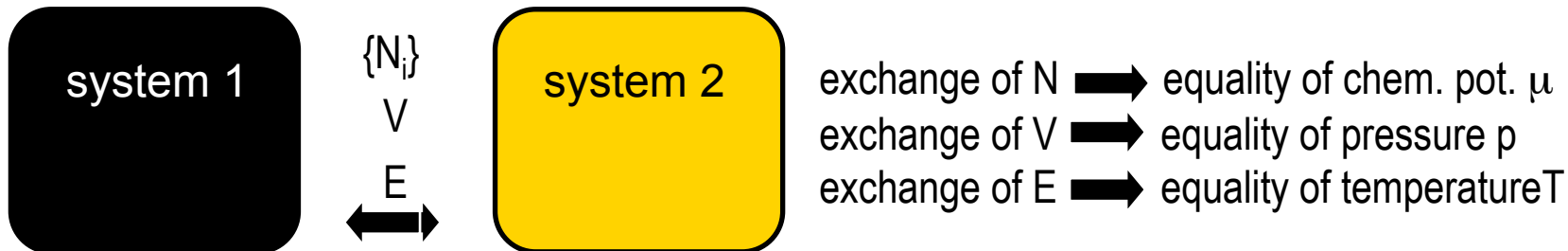
physical system is comprised of very many degrees of freedom,  $N_A \sim 10^{23}$   
 description of the microscopic state impractical but simple relations emerge  
 for thermodynamic variables in the limit  $N \rightarrow \infty$  (**thermodynamic limit**)

description of a physical system in terms of a **small number of state variables**

extensive variables: number of particles  $\{N_i\}$ , volume  $V$ , and internal energy  $E$

**(0. postulate)**

**fundamental question:** consider systems that can exchange properties



$$dU = -pdV + TdS + \sum_i \mu_i dN_i \quad \text{energy conservation}$$

**(1. postulate)**

reversible transformation      equality of entropy,  $S$  (i.e., reversible process,  $\Delta S=0$ )

**entropy** is additive,  $S(N, V, E) \rightarrow \max$  in equilibrium

**(2. postulate)**

$S$  monotonously increasing in  $E$ ,  $\partial S / \partial E|_{N,V} = \frac{1}{T} > 0$        $E(N, V, S) \rightarrow \min$

$$\partial E / \partial S|_{N,V} = 0 \Rightarrow S = 0 \quad (\text{Planck})$$

**(3. postulate)**

other ensembles via **Legendre transform:**

$$F(N, V, T) = E - TS$$

$$T = \left. \frac{\partial E}{\partial S} \right|_{N,V}$$

# basics: statistical mechanics (classical)

calculate thermodynamic properties for microscopic model

consider Hamiltonian  $\mathcal{H}(x_i, p_i) = \sum_i \frac{p_i^2}{2m} + \sum_{i < j} V(q_i - q_j)$

motion in phase space  $\Gamma$  on energy hypersurface,  $\mathcal{H}(x_i, p_i) = E$

calculate the time average of a quantity  $A(x_i, p_i)$  along trajectory

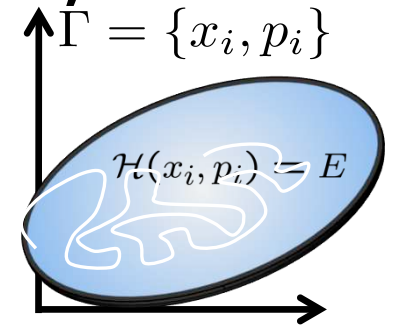
$$\begin{aligned} \bar{A}(\Gamma_0) &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T dt A(\Gamma(t, \Gamma_0)) \\ &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T dt \int_{\Gamma} d\Gamma' A(\Gamma') \delta[\Gamma' - \Gamma(t, \Gamma_0)] \\ &= \int_{\Gamma} d\Gamma' A(\Gamma') \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T dt \delta[\Gamma' - \Gamma(t, \Gamma_0)] \\ &= \frac{1}{N! h^{3N}} \int_{\Gamma} d\Gamma' A(\Gamma') \rho_{\text{mc}}(\Gamma') \quad (\text{ergodicity}) \end{aligned}$$

phase space density of microcanonical ensemble  $\rho_{\text{mc}} = \frac{1}{\mathcal{Z}_{\text{mc}}(N, V, E)}$   
is constant on hypersurface  $\mathcal{H}(x_i, p_i) = E$  and vanishes elsewhere

connection to thermodynamics via **Boltzmann's entropy functional**

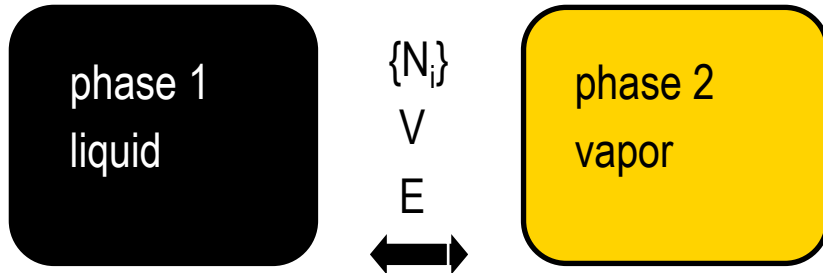
$$S[\rho] = -k_B \frac{1}{N! h^{3N}} \int d\Gamma \rho(\Gamma) \ln \rho(\Gamma) \longrightarrow S(N, V, E) = -k_B \langle \ln \rho_{\text{mc}}(\Gamma) \rangle = k_B \ln \mathcal{Z}_{\text{mc}}$$

other ensembles:  $F(N, V, T) = -k_B T \ln \mathcal{Z}$  with  $\mathcal{Z} = \frac{1}{N! h^{3N}} \int d\Gamma \exp\left(-\frac{E(\Gamma)}{k_B T}\right)$



# phase coexistence and transitions

example: liquid-vapor *coexistence*



**coexistence rule:** two phases coexist if they have the same pressure, temperature and chemical potential

grandcanonical ensemble  $\Omega(\mu, V, T) = -pV$   
 describe the phases by two thermodynamic potentials,  $\Omega_{\text{liq}}(\mu, V, T)$  and  $\Omega_{\text{vap}}(\mu, V, T)$

➡ coexistence:  $\Omega_{\text{liq}}(\mu, V, T) = \Omega_{\text{vap}}(\mu, V, T)$

**phase transition** (Ehrenfest)

first order: 1<sup>st</sup> derivative of thermodyn. potential (state variable)

is not continuous

➡ miscibility gap

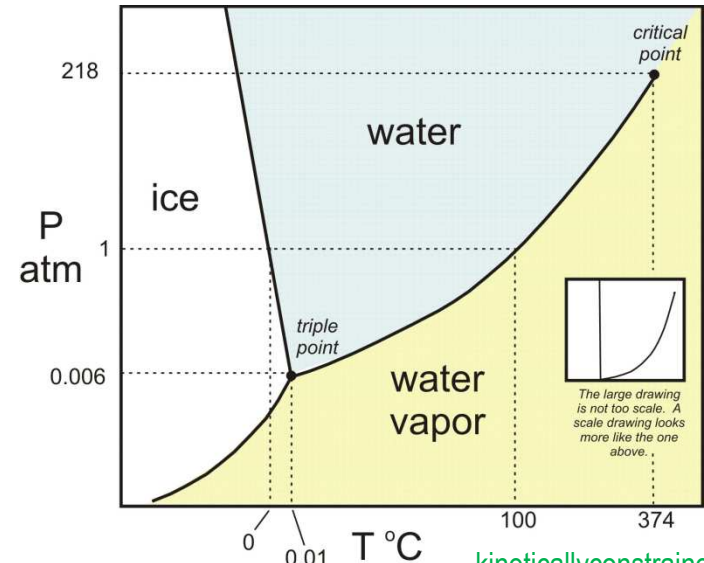
$$N = - \left. \frac{\partial \Omega}{\partial \mu} \right|_{V, T}$$

second order: 2<sup>nd</sup> derivative of TP (response function)

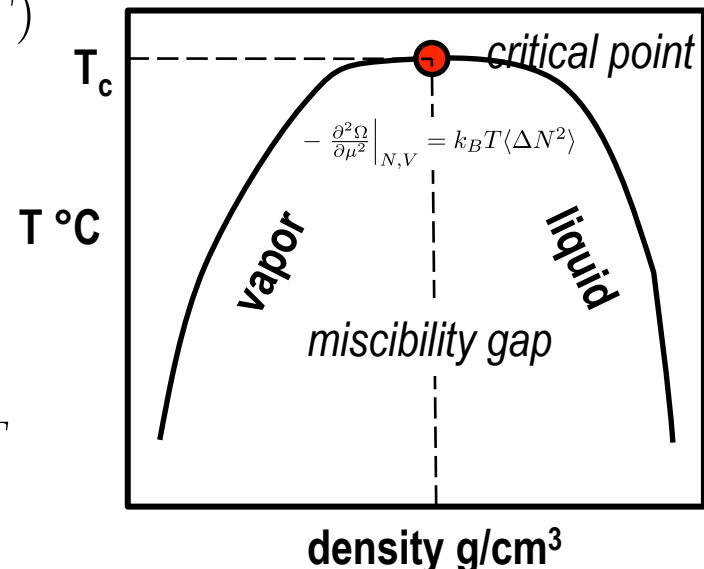
is not continuous

$$- \left. \frac{\partial^2 \Omega}{\partial \mu^2} \right|_{N, V} = \frac{\partial N}{\partial \mu} = \frac{N^2}{V} \kappa_T$$

➡ compressibility diverges at critical point



[kineticallyconstrained.blogspot.com/2009/05/critical-point.html](http://kineticallyconstrained.blogspot.com/2009/05/critical-point.html)





# basics: Monte-Carlo simulation

canonical ensemble (N,V,T) and thermal averages

$$\mathcal{Z} = \int \frac{\prod_{i=1}^N dp dr}{N! h^{3N}} \exp\left(-\frac{\mathcal{H}(\{p, r\})}{k_B T}\right)$$

$$\frac{1}{\Lambda_T} = \frac{1}{h} \int dp e^{-\frac{p^2}{2mk_B T}}$$

$$= \int \frac{\prod_{i=1}^N dr}{N! \Lambda_T^{3N}} \exp\left(-\frac{V(\{r\})}{k_B T}\right)$$

thermal de Broglie wavelength

$$\rho(\{r\}) N! \Lambda_T^{3N} = \frac{1}{\mathcal{Z}} \exp\left(-\frac{V(\{r\})}{k_B T}\right)$$

sharply peaked distribution in high-dimensional space

**idea:** generate sequence of points in phase space or configuration space that are distributed according to the statistical weight of the appropriate ensemble (**importance sampling**)

$$\langle A \rangle = \int \prod_{i=1}^N \rho(\{r\}) A(\{r\})$$

$$= \lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=1}^T A(\{r\}_t)$$

generate the sequence of configurations  $\{r\}_t$  as a **Markov chain**, i.e. the probability of generating the next configuration only depends on the previous state but not on the history

➡ choose the transition probability from one state to the next  $\{r\} \rightarrow \{r'\}$  as to generate  $\rho$

# basics: Monte-Carlo simulation

**master equation:** continuity equation for probabilities

$$\frac{\partial \rho(\{r\})}{\partial t} = + \underbrace{\sum_{\{r'\}} \rho(\{r'\}) p(\{r'\} \rightarrow \{r\})}_{\text{transitions into } \{r\}} - \underbrace{\rho(\{r\}) \sum_{\{r'\}} p(\{r\} \rightarrow \{r'\})}_{\text{transitions out of } \{r\}}$$

stationary  $\frac{\partial \rho_{\text{eq}}(\{r\})}{\partial t} = 0$

$$\sum_{\{r'\}} \rho_{\text{eq}}(\{r'\}) p(\{r'\} \rightarrow \{r\}) = \rho_{\text{eq}}(\{r\}) \sum_{\{r'\}} p(\{r\} \rightarrow \{r'\})$$

**detailed balance:** reversibility between all pairs of microstates (sufficient condition)

$$\rho_{\text{eq}}(\{r'\}) p(\{r'\} \rightarrow \{r\}) = \rho_{\text{eq}}(\{r\}) p(\{r\} \rightarrow \{r'\}) \quad \forall \{r\} \leftrightarrow \{r'\}$$

decompose the transition probability into a proposal and acceptance probability

$$p(\{r\} \rightarrow \{r'\}) = p_{\text{prop}}(\{r\} \rightarrow \{r'\}) p_{\text{acc}}(\{r\} \rightarrow \{r'\})$$

$$\frac{p_{\text{acc}}(\{r\} \rightarrow \{r'\})}{p_{\text{acc}}(\{r'\} \rightarrow \{r\})} = \frac{p_{\text{prop}}(\{r'\} \rightarrow \{r\}) p_{\text{eq}}(\{r'\})}{p_{\text{prop}}(\{r\} \rightarrow \{r'\}) p_{\text{eq}}(\{r\})}$$

$$\longrightarrow p_{\text{acc}}(\{r\} \rightarrow \{r'\}) = \text{metrop} \left[ \frac{p_{\text{prop}}(\{r'\} \rightarrow \{r\}) p_{\text{eq}}(\{r'\})}{p_{\text{prop}}(\{r\} \rightarrow \{r'\}) p_{\text{eq}}(\{r\})} \right]$$

with  $\text{metrop}[x] = \min(1, x)$  and  $\frac{\text{metrop}[x]}{\text{metrop}[1/x]} = x$

# basics: Monte-Carlo simulation

## advantages of Monte-Carlo sampling:

- ◆ fast exploration of phase space (configuration space) by moves that efficiently renew configurations (non-local Monte-Carlo moves like (re)growing part of a macromolecule or all-chemical mutations)
- ◆ can be formulated in various ensembles (e.g., moves that insert or remove particles, change volume)
- ◆ a priori no information about the dynamics but “sometimes” can mimic the dynamics of molecules

example: force-bias Monte-Carlo move or Smart-Monte-Carlo move ([Rosky, Doll, Friedman, 1978](#))

**idea:** use Brownian dynamics with a large time step to propose a move

$$\mathbf{r}_i(t + dt) - \mathbf{r}_i(t) = \frac{dt}{\zeta} \mathbf{F}_i(t) + \sqrt{2k_B T \frac{dt}{\zeta}} \eta + \mathcal{O}(dt^2)$$

$$= dA \mathbf{F}_i(t) + dR \eta + \mathcal{O}(dt^2) \quad \text{with} \quad dA = \frac{dt}{\zeta} \quad dR = \sqrt{2k_B T dA}$$

$$\text{propose } \mathbf{r}_i \rightarrow \mathbf{r}'_i = \mathbf{r}_i + \Delta A \mathbf{F}_i + \Delta \mathbf{R}_i \quad \text{with} \quad P(\Delta R_i) = \frac{1}{\sqrt{4\pi k_B T \Delta A}} \exp \left[ -\frac{\Delta R_i^2}{4k_B T \Delta A} \right]$$

$$p_{\text{prop}}(\{r\} \rightarrow \{r'\}) = \frac{1}{(4\pi k_B T \Delta A)^{3/2}} \exp \left[ -\frac{(\mathbf{r}'_i - \mathbf{r}_i - F_i \Delta A)^2}{4k_B T \Delta A} \right]$$



$$p_{\text{acc}}(\{r\} \rightarrow \{r'\}) = \text{metrop} \left[ \exp \left( -\frac{V' - V}{k_B T} - \frac{\mathbf{F}'_i + \mathbf{F}_i}{2k_B T} (\mathbf{r}'_i - \mathbf{r}_i) - \frac{\Delta A}{4k_B T} \{ \mathbf{F}'_i{}^2 - \mathbf{F}_i{}^2 \} \right) \right]$$

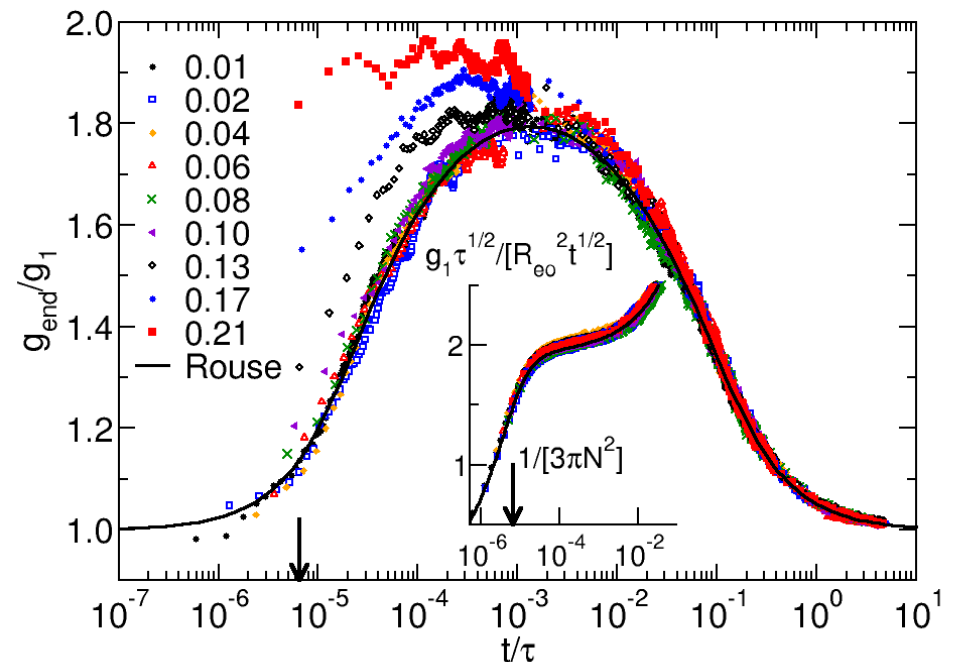
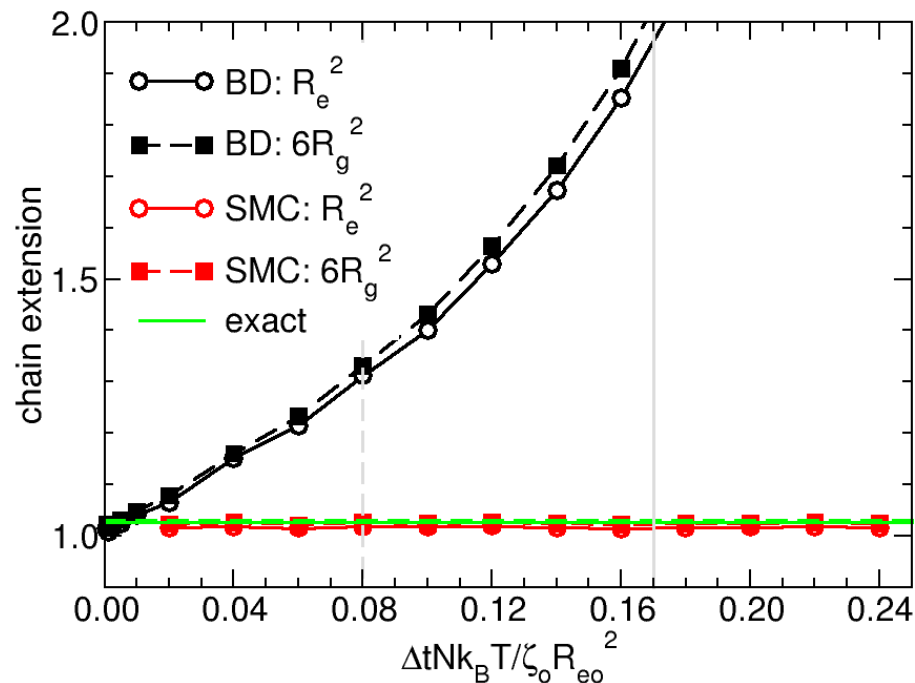
# Rouse-like dynamics via Smart-MC simulations

**SMC:** Brownian dynamics as smart-MC (Rosky, Doll, Friedman, 1978)

**idea:** uses forces to construct trial displacements  $\Delta r$

**application:** soft, coarse-grained polymer model

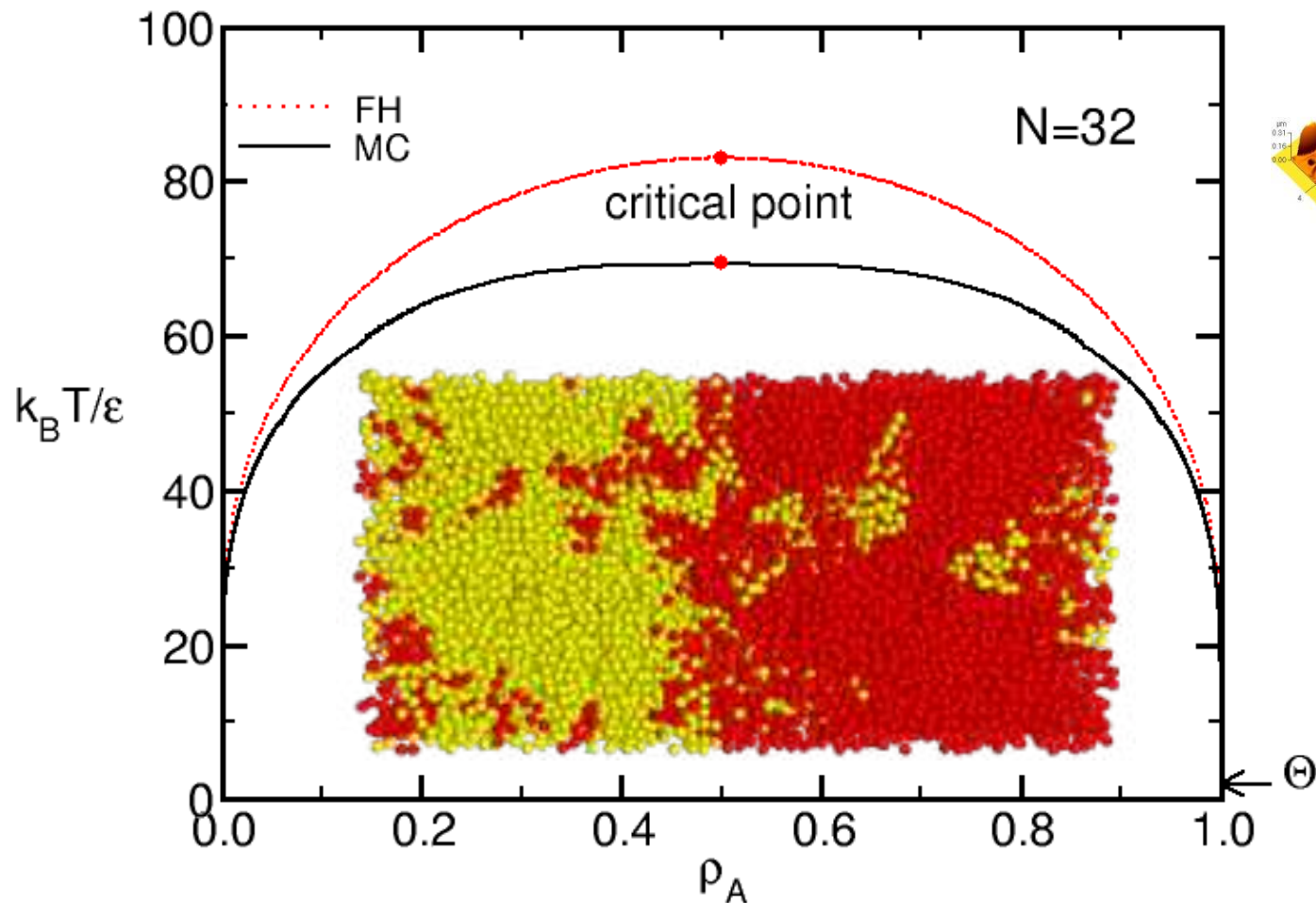
- SMC or force-bias MC allows for a larger time step (factor 100) than BD with Euler integration
- rather realistic dynamics (diffusive) after a few Smart-Monte-Carlo steps



# prototypical example: liquid-liquid phase separation in blends

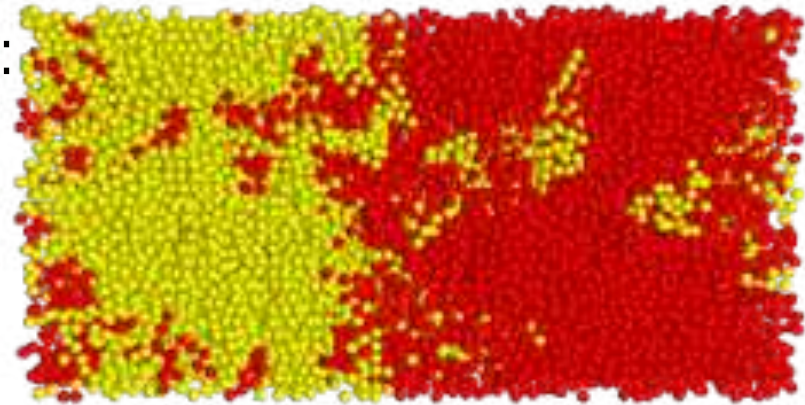
demixing into A-rich and B-rich regions, interfaces in a polymer blend

**order parameter:** composition difference between coexisting phases



# methods to locate phase coexistence (1)

1. **direct simulation** of phase coexistence:  
two phases simultaneously present in a simulation box separated by interfaces



advantage:

simple, does not require special simulation techniques, does not require priori knowledge of coexistence value of pressure or chemical potential

interface tension accessible via anisotropy of pressure across interface

$$\gamma = \int_{-\infty}^{\infty} dz [p_N(z) - p_T(z)]$$

disadvantage:

requires very large systems to distinguish interface and bulk properties

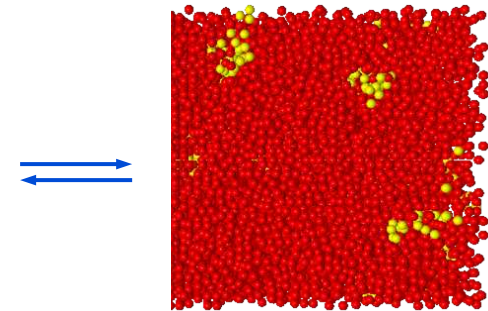
often kinetics of phase separation is very slow and equilibration difficult



# methods to locate phase coexistence (2)

## 2. Gibbs-Ensemble Monte-Carlo simulations:

two simulation boxes which  
exchange particles and volume  
to ensure that boxes are at constant  
chemical potential and pressure



advantage:

no interfaces present, does not require a priori knowledge of coexistence  
value of pressure or chemical potential

disadvantage:

requires additional MC moves for volume and particle exchange that  
might be difficult to implement (e.g., lattice models)

# methods to locate phase coexistence (3)

## (semi-) grandcanonical ensemble:

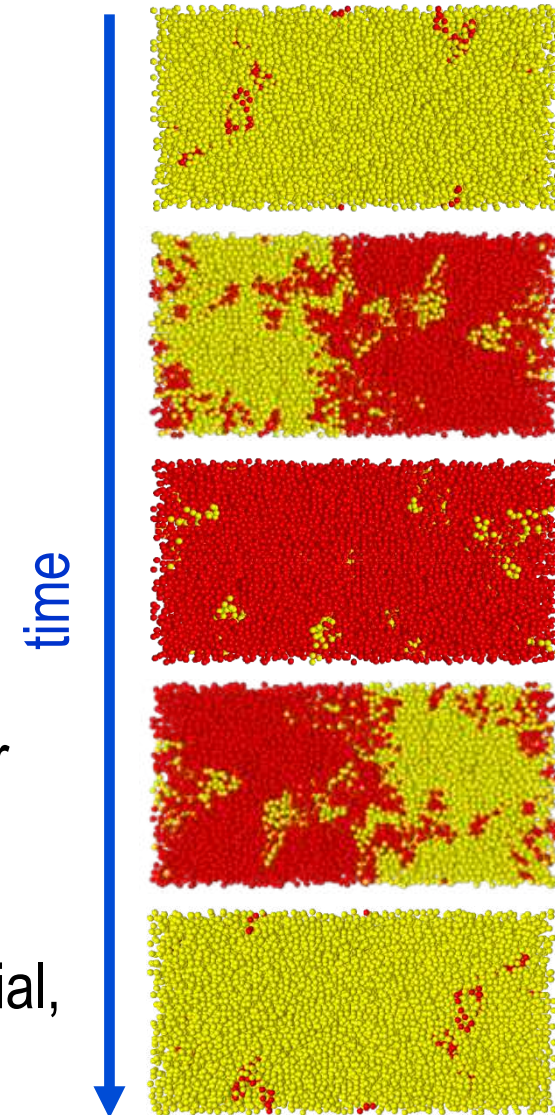
control temperature  $T$ , volume  $V$ , and total number of polymers, exchange potential,  $\Delta\mu$  (semi-gc) or chemical potential,  $\mu$  (grandcanonical)  
order parameter fluctuates

## advantage:

probability distribution of order parameter (density or composition) contains information about bulk phases and interfaces, efficient relaxation of order parameter

## disadvantage:

needs prior “guess” for coexistence chemical potential, special techniques to (i) change order parameter and (ii) efficiently sample the order parameter distribution



# phase coexistence and equal weight rule

## thermodynamics:

two phases coexist at  $\mu$  if they have the same grand potential

$$\begin{aligned}\Omega(\mu, T) &= -k_B T \ln \mathcal{Z}_{\text{gc}} \\ &= -pV\end{aligned}$$

## statistical mechanics:

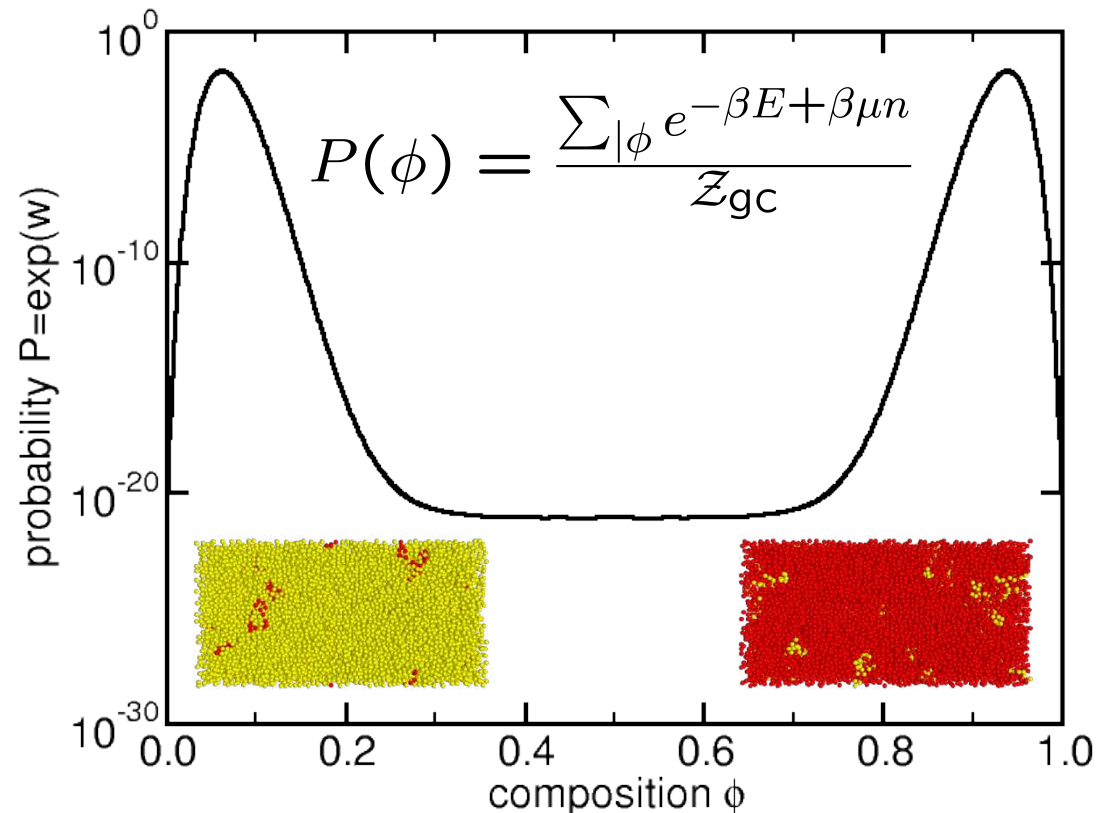
$$\mathcal{Z}_{\text{gc}} = \sum_{\{\text{conf}\}} e^{-\beta E + \beta \mu n}$$

$$\mathcal{Z}_A = \sum_{\{A\}} e^{-\beta E + \beta \mu n} = \mathcal{Z}_{\text{gc}} \int_A d\phi P(\phi)$$

$$\mathcal{Z}_B = \sum_{\{B\}} e^{-\beta E + \beta \mu n} = \mathcal{Z}_{\text{gc}} \int_B d\phi P(\phi)$$

$$\Omega_A \stackrel{!}{=} \Omega_B \quad \Rightarrow \quad \int_A d\phi P(\phi) = \int_B d\phi P(\phi)$$

➔ exponentially small finite-size effects



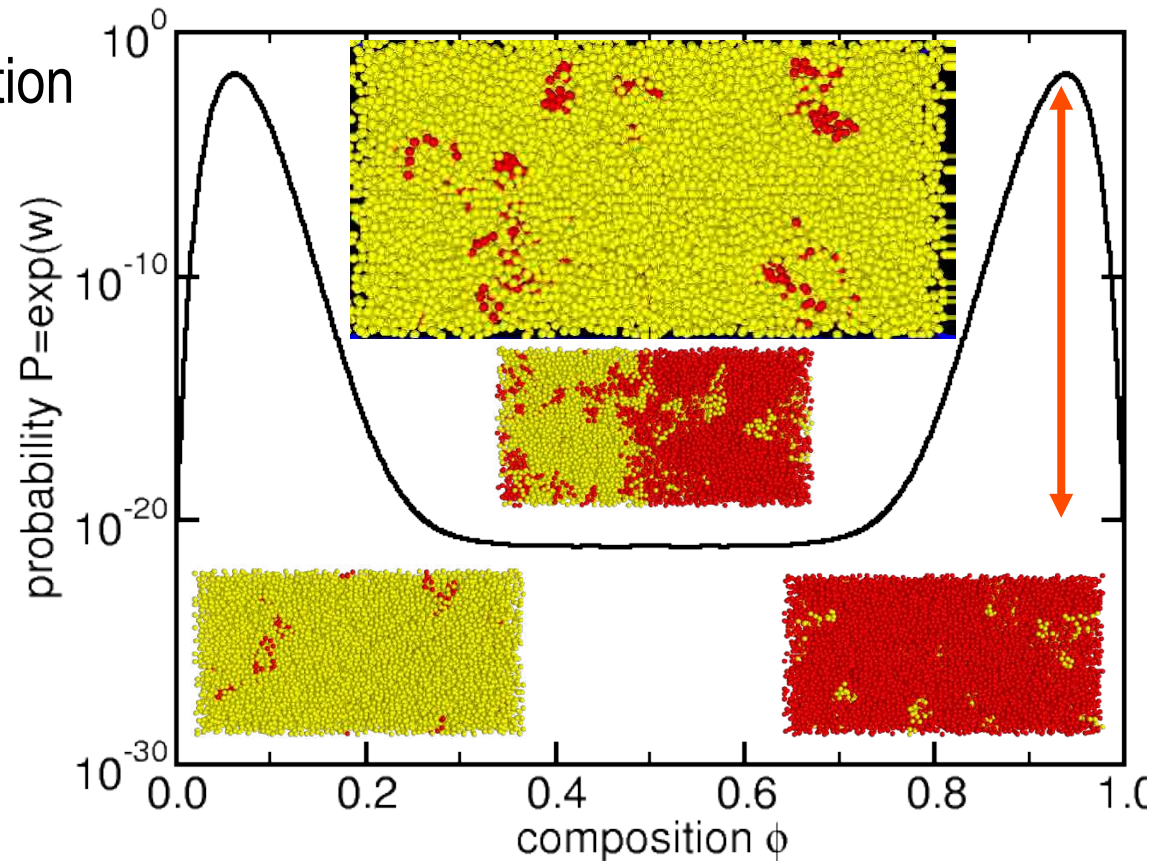
# semi-grandcanonical simulations

canonical moves:  
local displacements+reptation

grand-canonical identity  
switch:



- phase coexistence via equal weight rule
- interface properties via “depth of valley”



flat valley in order parameter indicates that interfaces do not interact  
use elongated simulation cell, displaces the interfaces further

# how to tunnel between the two phases?

**non-Boltzmann sampling**  
(reweighting)

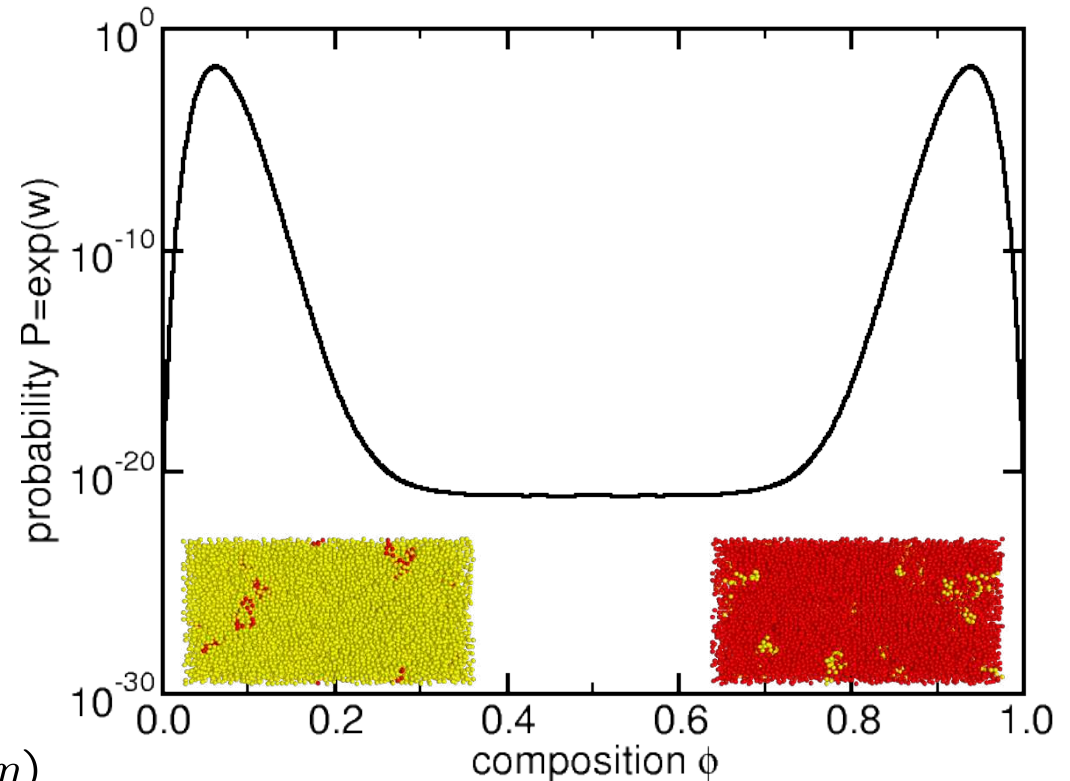
$$\mathcal{H}(\{r\}) = \mathcal{H}_0(\{r\}) + k_B T \eta(\phi)$$

additional term only depends on  
composition  $\phi$  but not on  
detailed configuration  $\{r\}$

**idea**  $\eta(\phi) = \ln P_{gc}$

$$\begin{aligned} P_{rw}(\phi) &\sim e^{-\beta(\mathcal{H}_0 - \mu n + k_B T \eta)} \\ &\approx e^{-\beta(\mathcal{H}_0 - \mu n)} / P \\ &\sim 1 \end{aligned}$$

➔ uniform sampling



relative probability of configurations with  
same  $n$  (order par.) remains unaffected

# estimating the reweighting function

## Wang-Landau algorithm

Wang, Landau PRL **86**, 2050 (2001)

adjust  $\eta$  every step

$$\eta \rightarrow \eta + \ln f$$

$$P_{\text{rw}}(\phi) \sim P(\phi)e^{-\eta(\phi)}$$

accumulate histogram

flat histogram?

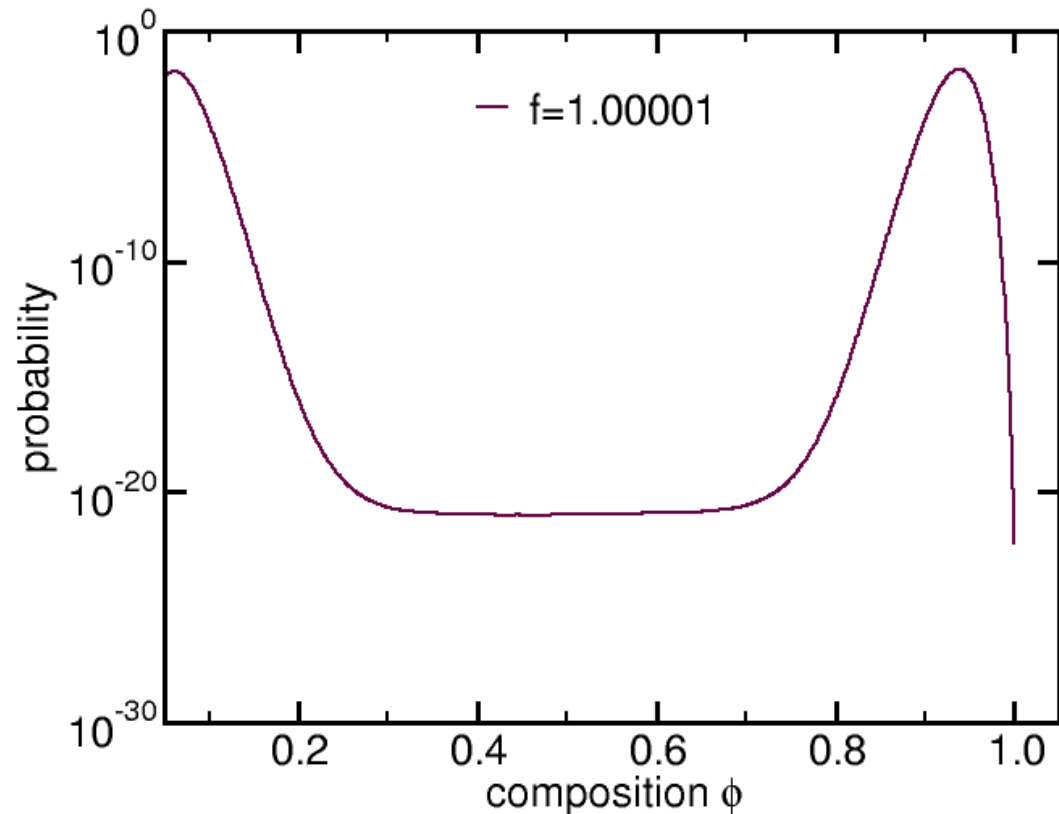
$$f \rightarrow \sqrt{f} \quad (\rightarrow 1)$$

reset histogram

but: no detailed balance for  $f > 1$

errors difficult to estimate

**idea:** push out the simulation of states that were already visited



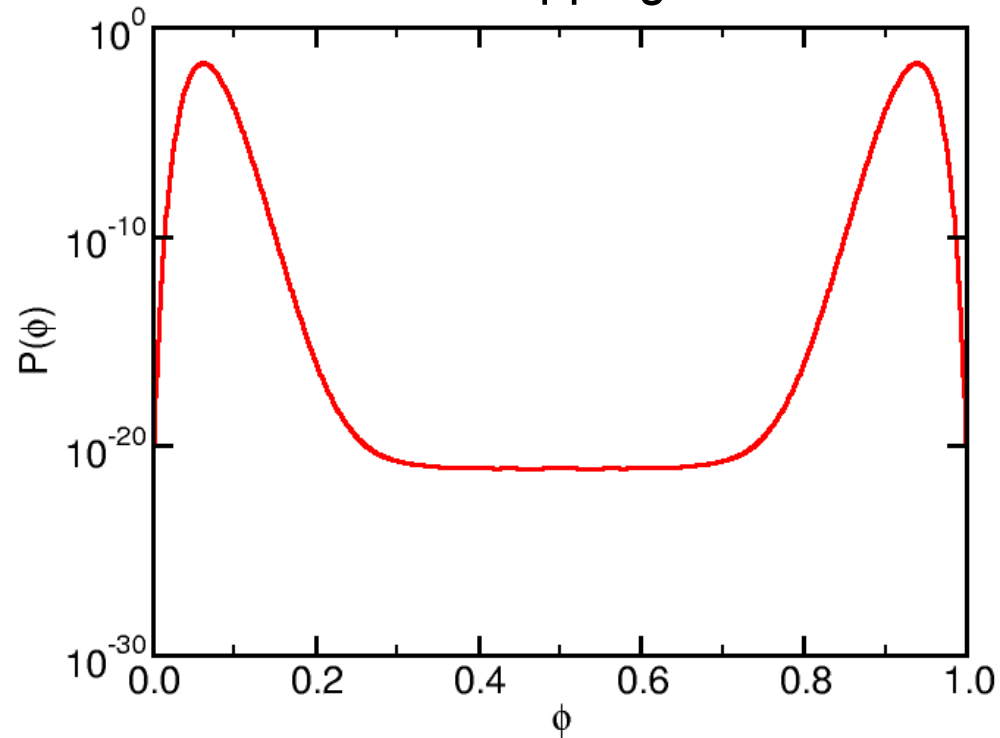


# umbrella sampling

divide the interval of particle numbers into overlapping subintervals

obtain  $P_i(\phi)$  in  
each subinterval

match distributions  
at boundaries



window size:

- smaller windows are easier to sample because weight does not vary
- larger windows might allow for additional relaxation modes and prevent sampling difficulties

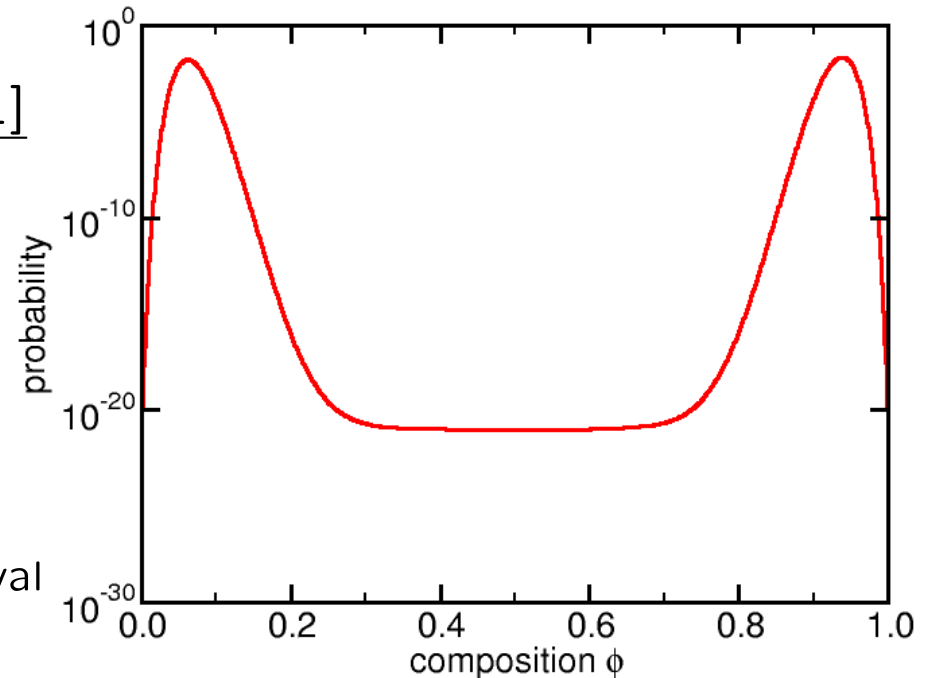
# successive umbrella sampling

simulate intervals  $[j, j+1]$  of A-polymers

$$P[i] \sim \prod_{j < i} \frac{P[j+1]}{P[j]} \approx \prod_{j < i} \frac{H[j+1]}{H[j]}$$

error:

$$(\Delta P[n])^2 \sim n \left( \Delta \frac{H[j+1]}{H[j]} \right)^2 \sim n \underbrace{\frac{n}{\#MCS}}_{\text{time spent in interval}}$$



- in the absence of sampling difficulties, similar to a simulation of complete interval with perfect reweighting

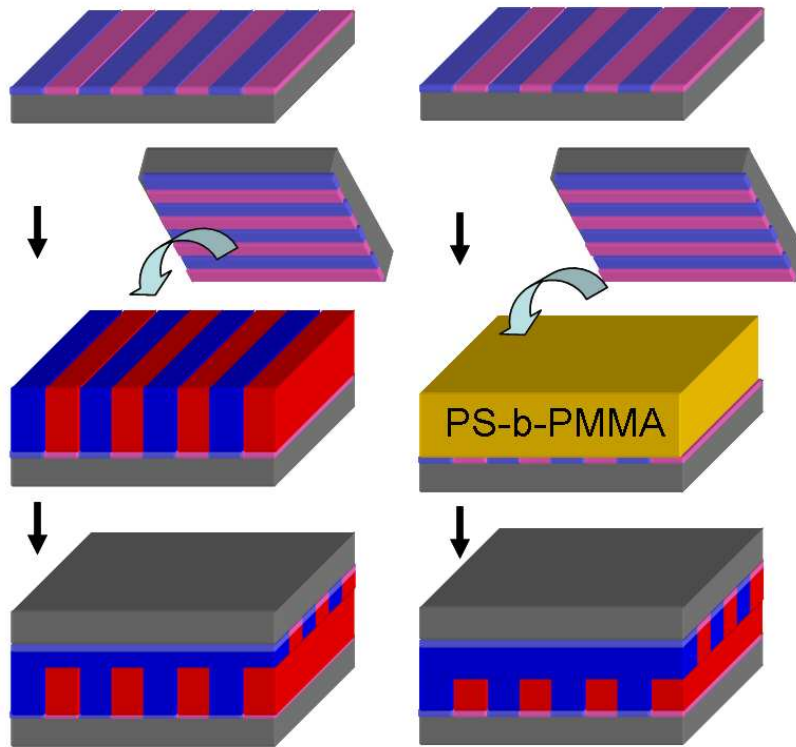
$$(\Delta P[n])^2 \sim \frac{1}{\text{tunnellings } 1 \leftrightarrow n} \sim \frac{n^2}{\#MCS}$$

- error allows for identification of sampling errors

➡ total error does not depend on the window size

# interface localization delocalization-transition

morphology between opposing stripe patterns



boundaries stabilize a twist grain boundary, which resembles Scherk's first surface

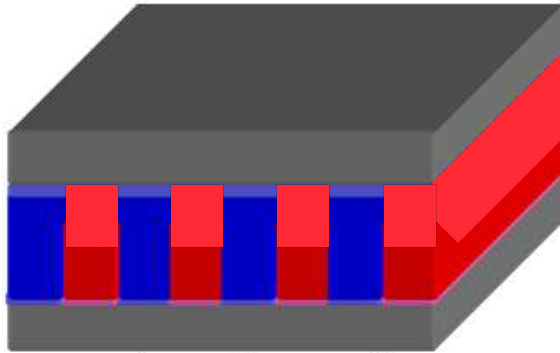
Liu, Ramirez-Hernandez, Yoshida, Nygard, Satapathy, Bunk, de Pablo, Nealey, PRL 108, 065502 (2012)

**question:** location of the grain boundary ?

➔ control of three-dimensional morphology

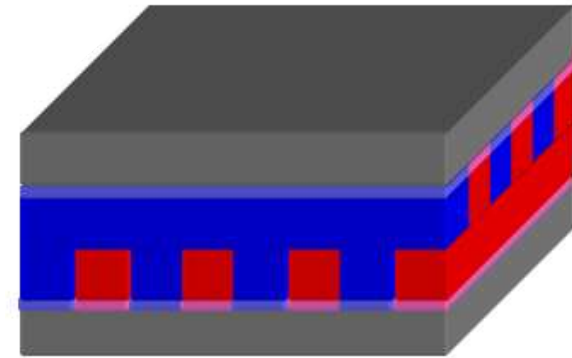
analogy between behavior of twist grain boundary and wetting in a film with strictly anti-symmetric boundaries (interface localization-delocalization)  
control via geometric parameters of the system

# interface localization delocalization-transition



## localized state:

twist-grain boundary localized close top or bottom surface (partial wetting)



## delocalized state:

twist-grain boundary fluctuates around the center of the film (repelled by both surfaces, complete wetting)

➡ use position,  $x$ , of grain boundary as order parameter

mesoscopic free energy:

$$\frac{\Delta F(x)}{A} = \gamma_b(\Lambda) + \gamma_t(\Lambda) + \gamma(\alpha) \quad \text{surface and interface tensions (Young)}$$

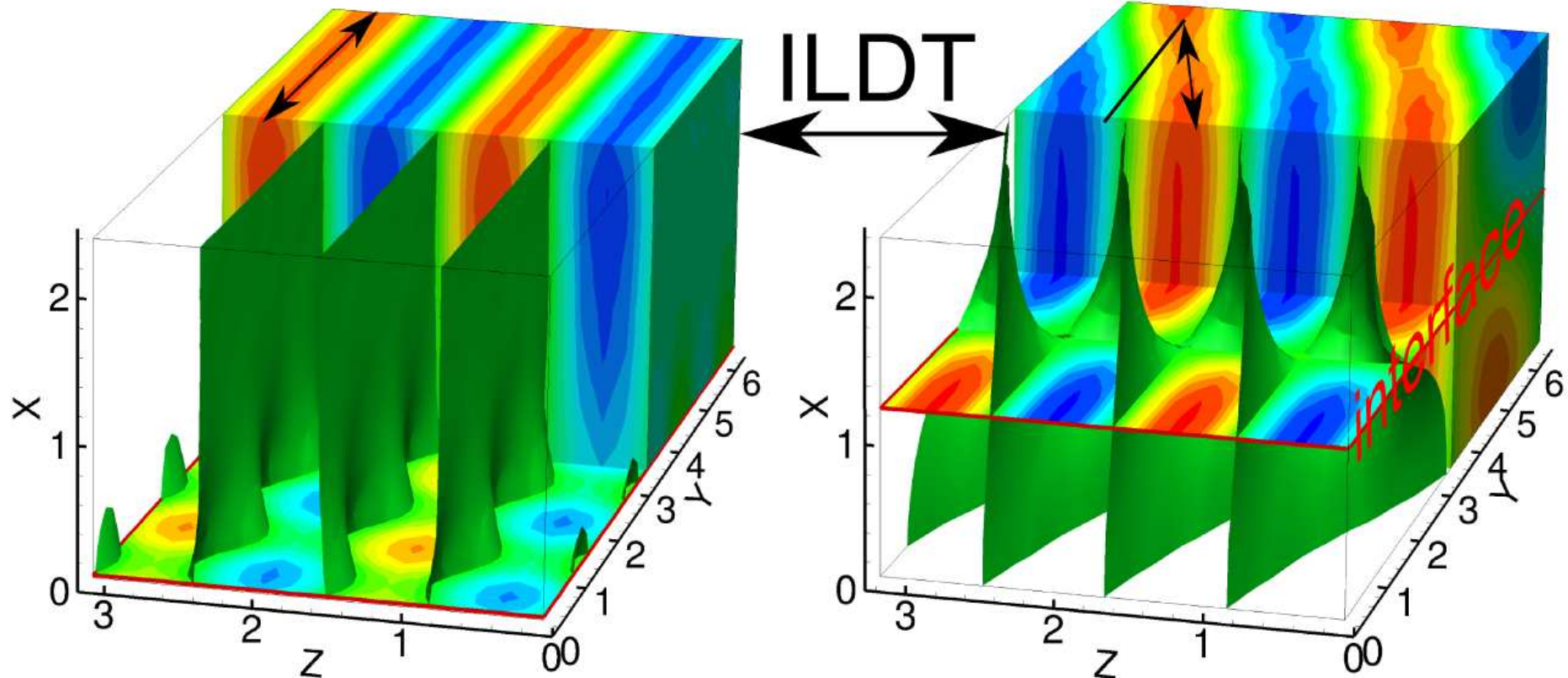
$$+ \frac{1}{2}B(\lambda_b - \lambda_0)^2 x + \frac{1}{2}B(\lambda_t - \lambda_0)^2 (D - x) \quad \text{stretch/compression } (\mu)$$

$$+ g_b(x) + g_t(D - x) \quad \text{short-range interface potential due to distortion}$$

# interface localization delocalization-transition

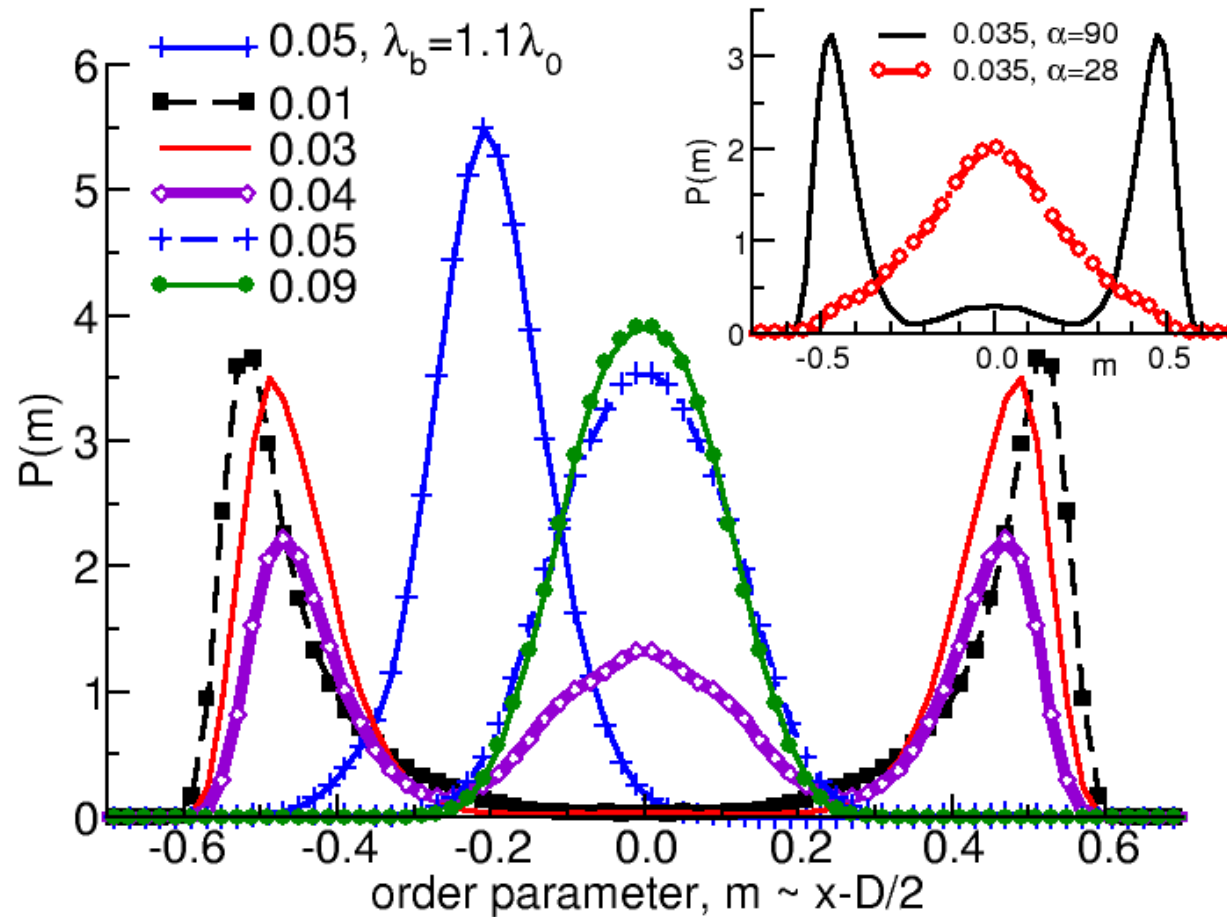
twist angle  $\alpha=90^\circ$ , localized at bottom

twist angle  $\alpha=28^\circ$ , delocalized



$$\begin{aligned} \frac{\Delta F(x)}{A} &= \gamma_b(\Lambda) + \gamma_t(\Lambda) + \gamma(\alpha) \\ &+ \frac{1}{2}B(\lambda_b - \lambda_0)^2 x + \frac{1}{2}B(\lambda_t - \lambda_0)^2 (D - x) \\ &+ g_b(x) + g_t(D - x) \end{aligned}$$

# interface localization delocalization-transition



$$\frac{\Delta F(x)}{A} = \gamma_b(\Lambda) + \gamma_t(\Lambda) + \gamma(\alpha)$$

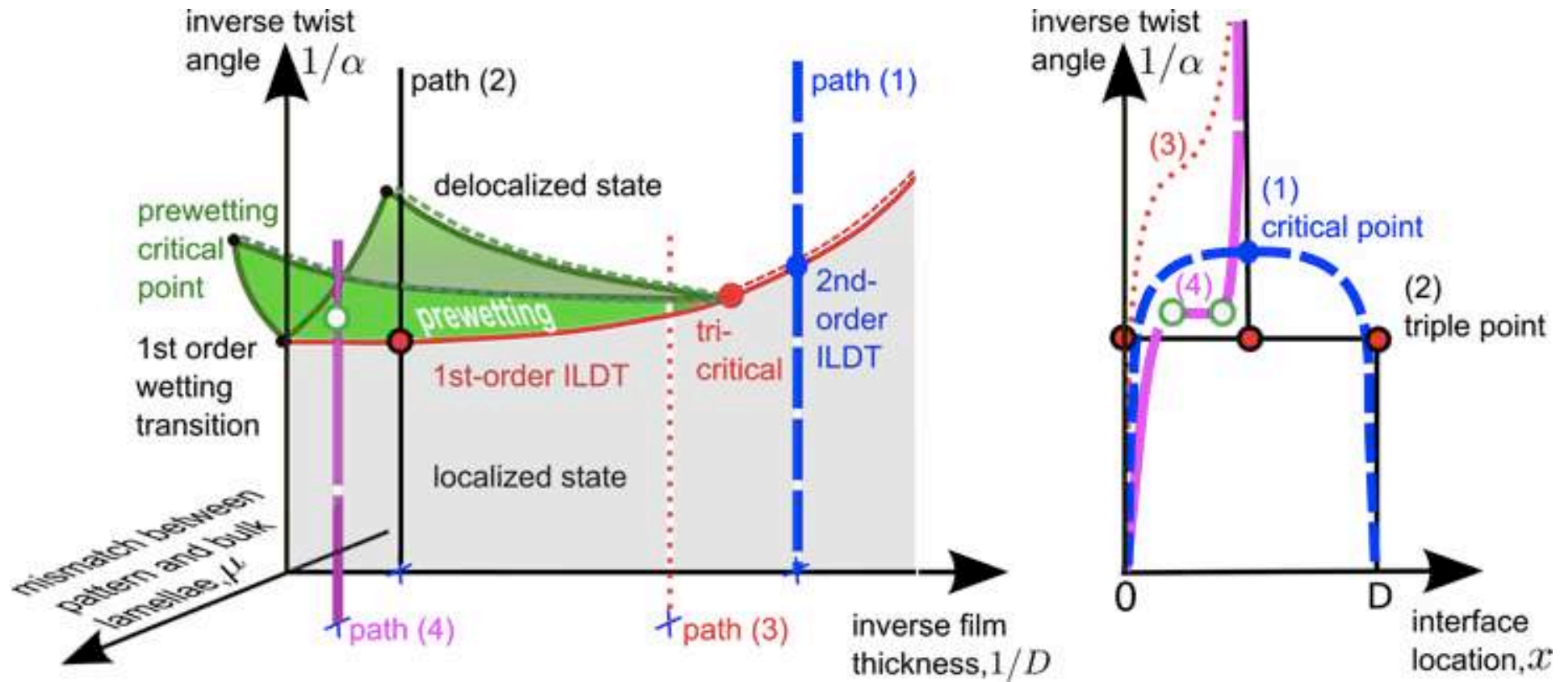
$$D = 2.468R_{e0}, L_y = 2L_z = 2\lambda_0$$

$$+ \frac{1}{2}B(\lambda_b - \lambda_0)^2 x + \frac{1}{2}B(\lambda_t - \lambda_0)^2 (D - x)$$

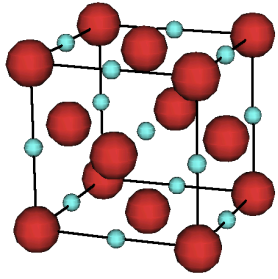
$$+ g_b(x) + g_t(D - x)$$



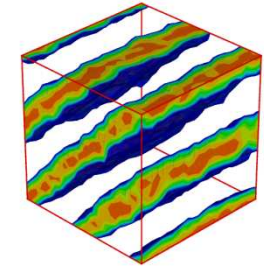
# interface localization delocalization-transition



- analogy between behavior of twist grain boundary and wetting in a film with strictly anti-symmetric boundaries (interface localization-delocalization)
- control via geometric parameters of the system ( $T \sim 1/\alpha$  and  $\mu \sim (\lambda - \lambda_0)^2$ )



# crystallization vs self-assembly



## **order parameter:**

Fourier mode of density fluctuation  
ideal ordered state: ideal crystal ( $T=0$ )  
disordered state: ideal gas

ordered state: particles vibrate  
around ideal lattice positions

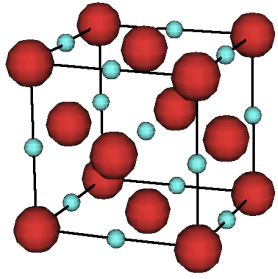
Einstein crystal is reference state  
use thermodynamic integration wrt  
to uniform, harmonic coupling of  
particles to ideal position

(Frenkel & Ladd)

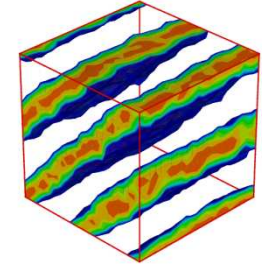
Fourier mode of composition fluctuation  
SCFT solution  
homogeneous fluid/melt

ordered phase: composition fluctuates  
around reference state (SCFT solution),  
but **molecules diffuse** (liquid)

**no simple reference state** for  
self-assembled morphology



# crystallization vs self-assembly



## **order parameter:**

Fourier mode of density fluctuation  
ideal ordered state: ideal crystal ( $T=0$ )  
disordered state: ideal gas

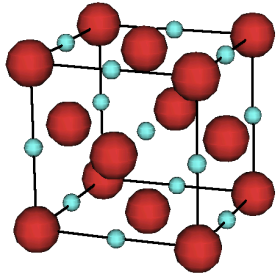
ordered state: particles vibrate  
around ideal lattice positions

Fourier mode of composition fluctuation  
SCFT solution  
homogeneous fluid/melt

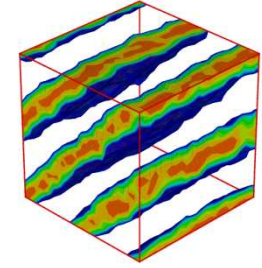
ordered phase: composition fluctuates  
around reference state (SCFT solution),  
but **molecules diffuse** (liquid)

free energy per molecule  $N k_B T$   
relevant free-energy differences  $10^{-3} k_B T$

➡ absolute free energy must be measured with a relative accuracy of  $10^{-5}$



# crystallization vs self-assembly



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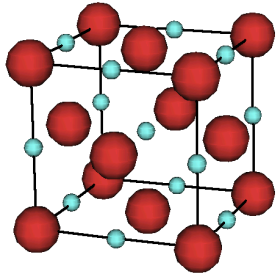
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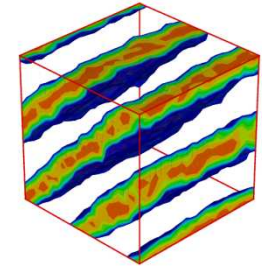
ordered phase: composition fluctuates  
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but ***molecules diffuse*** (liquid)

free energy per molecule  $N k_B T$   
relevant free-energy differences  $10^{-3} k_B T$

➡ absolute free energy must be measured with a relative accuracy of  $10^{-5}$   
measure free energy differences by **reversibly transforming one structure into another** ( $10^{-3}$  relative accuracy needed)



# crystallization vs self-assembly



## **order parameter:**

Fourier mode of density fluctuation  
ideal ordered state: ideal crystal ( $T=0$ )  
disordered state: ideal gas

ordered state: particles vibrate  
around ideal lattice positions

Fourier mode of composition fluctuation  
SCFT solution  
homogeneous fluid/melt

ordered phase: composition fluctuates  
around reference state (SCFT solution),  
but **molecules diffuse** (liquid)

**How a solid can be turned into a gas without passing through a first-order phase transformation**

Sheh-Yi Sheu and Chung-Yuan Mou

*Department of Chemistry, National Taiwan University, Taipei, Taiwan, Republic of China*

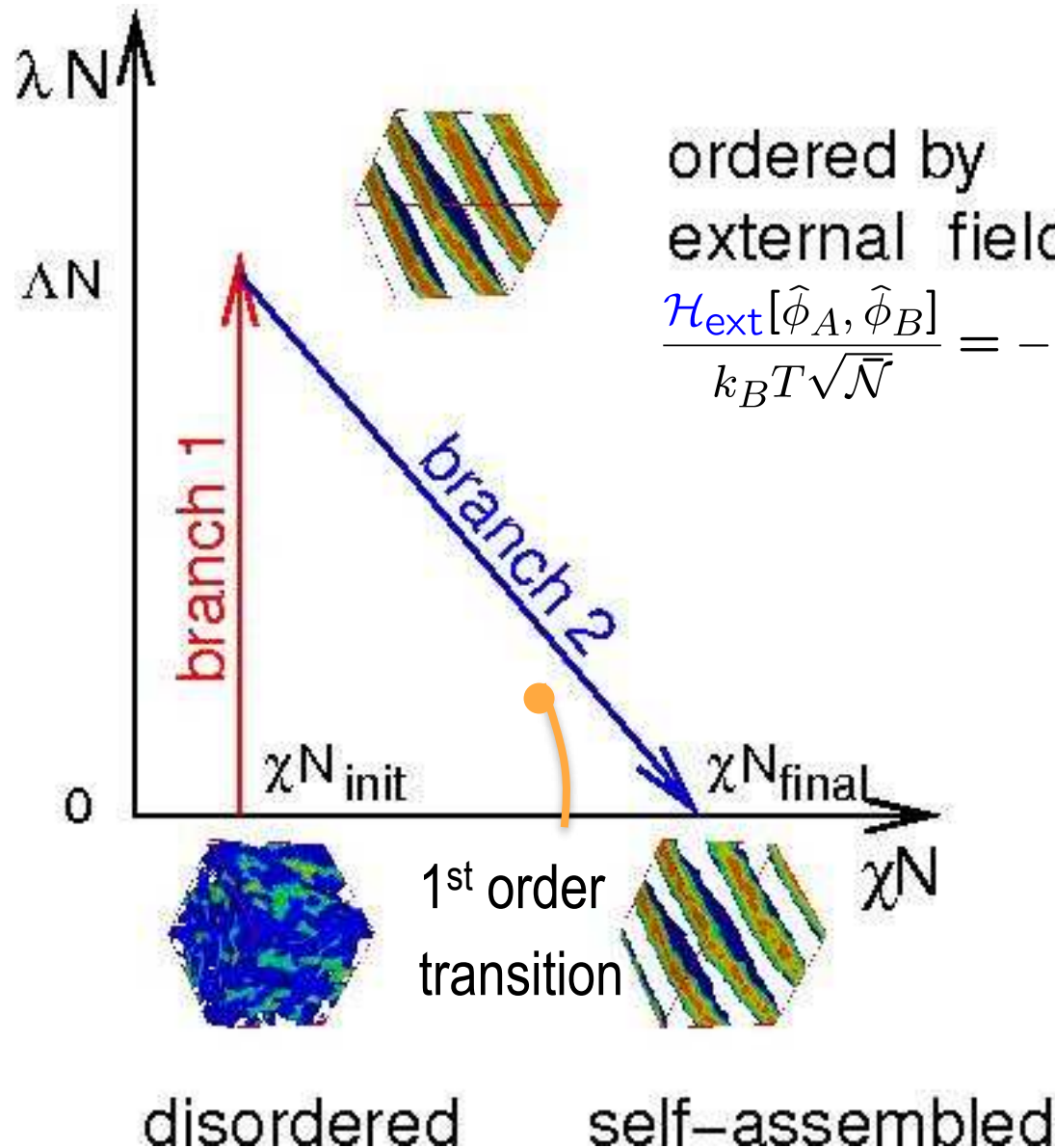
PRE 51, R3795 (1995)

Ronald Lovett

*Department of Chemistry, Washington University, St. Louis, Missouri 63130*

see also Grochola, JCP 120, 2122 (2004)

# calculating free energy differences



ordered by external field

$$\frac{\mathcal{H}_{ext}[\hat{\phi}_A, \hat{\phi}_B]}{k_B T \sqrt{\bar{N}}} = -\lambda N \int \frac{d^3 \mathbf{r}}{R_{e0}^3} h(\mathbf{r}) [\hat{\phi}_A(\mathbf{r}) - \hat{\phi}_B(\mathbf{r})]$$

intermediate state:  
independent chains in static, external field (SCFT)

## branch 1:

“non-interacting  
= no collective phenomena”

## branch 2:

ideally: no structural change  
condition for ordering field

$$\bar{N} = 14\,884, V = (4.88 R_e)^3$$

$$n = 13\,241, N = 32$$

$$\Delta L = 0.19675 R_e$$

Müller, Daoulas, JCP 128, 024903 (2008)



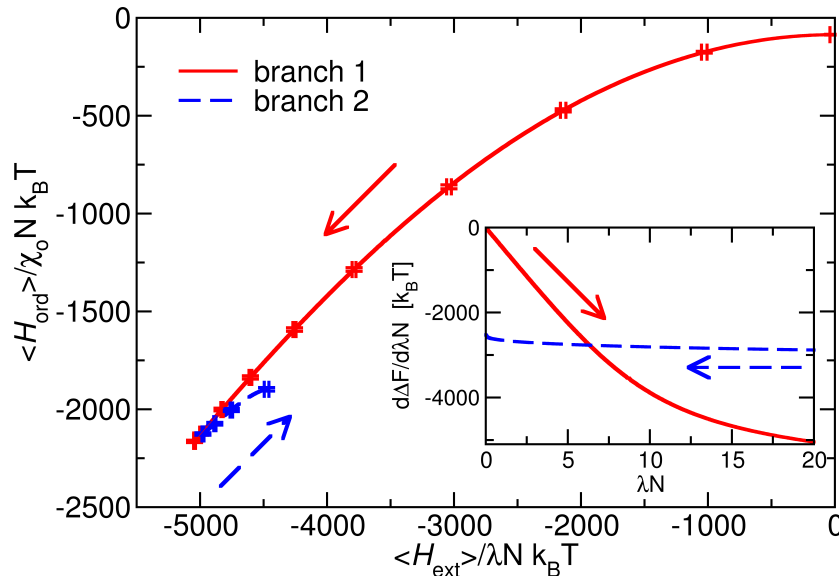
# free energy difference via TDI

$$\mathcal{H}_{\text{nb}} = \mathcal{H}_{\text{liq}} + \mathcal{H}_{\text{ord}} + \mathcal{H}_{\text{ext}} \quad \frac{\mathcal{H}_{\text{ord}}}{nk_B T} = -\chi_0 N f_{\text{ord}}[\phi] \quad \text{with } \phi = \phi_A - \phi_B$$

$$\frac{\mathcal{H}_{\text{ext}}}{nk_B T} = -\frac{\lambda N}{V} \int_V d^3\mathbf{r} h(\mathbf{r})\phi(\mathbf{r})$$

$$F = -k_B T \ln \int \mathcal{D}[\{r\}] \exp \left[ -\frac{\mathcal{H}_b + \mathcal{H}_{\text{liq}} + \mathcal{H}_{\text{ord}} + \mathcal{H}_{\text{ext}}}{k_B T} \right]$$

$$\frac{\partial F}{\partial \lambda N} = -k_B T \frac{\int \mathcal{D}[\{r\}] - \frac{1}{k_B T} \frac{\partial \mathcal{H}_{\text{ext}}}{\partial \lambda N} \exp \left[ -\frac{\mathcal{H}_b + \mathcal{H}_{\text{liq}} + \mathcal{H}_{\text{ord}} + \mathcal{H}_{\text{ext}}}{k_B T} \right]}{\int \mathcal{D}[\{r\}] \exp \left[ -\frac{\mathcal{H}_b + \mathcal{H}_{\text{liq}} + \mathcal{H}_{\text{ord}} + \mathcal{H}_{\text{ext}}}{k_B T} \right]} = \left\langle \frac{\mathcal{H}_{\text{ext}}}{\lambda N} \right\rangle$$



optimal choice of external field (Sheu et al):  
structure does not change along 2<sup>nd</sup> branch

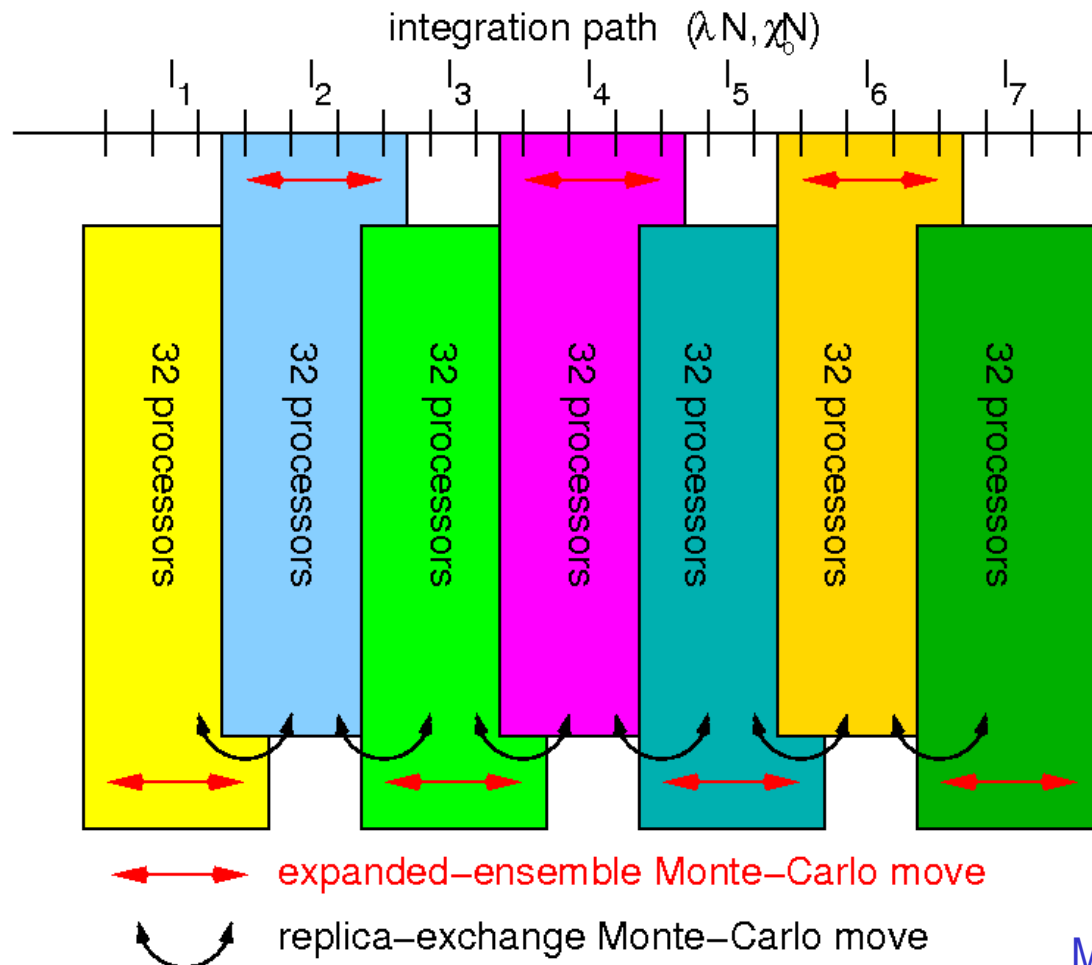
$$-\lambda N h_{\text{eff}}(\mathbf{r}) = \frac{V}{nk_B T} \left( \frac{\delta \mathcal{H}_{\text{ord}}}{\delta \phi(\mathbf{r})} + \frac{\delta \mathcal{H}_{\text{ext}}}{\delta \phi(\mathbf{r})} \right)$$

$$= -\lambda N h(\mathbf{r}) - \chi_0 N \frac{\delta f_{\text{ord}}}{\delta \phi(\mathbf{r})}$$

$$\text{SCFT: } \chi_0 N(\lambda N) = \chi_0 N(0) - 2\lambda N$$

# TDI vs expanded ensemble/replica exchange

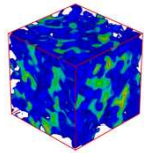
$$\mathcal{Z} \sim \prod_{j=1}^{M_r} \sum_{\lambda_i N \in I_j} \frac{e^{\eta(\lambda_i N)}}{n!} \int \prod_{i=1}^n \mathcal{D}[\mathbf{r}_i(s)] e^{-\frac{\mathcal{H}_b + \mathcal{H}_{\text{liq}} + \mathcal{H}_{\text{ord}} + \mathcal{H}_{\text{ext}}}{k_B T}}$$



only replica exchange is impractical because one would need several 100 configurations at initial stage, where weights are unknown ( $\Delta F \sim 10^4 k_B T$ ), replica exchange guarantees more uniform sampling expanded ensemble technique is useful because it provides an error estimate

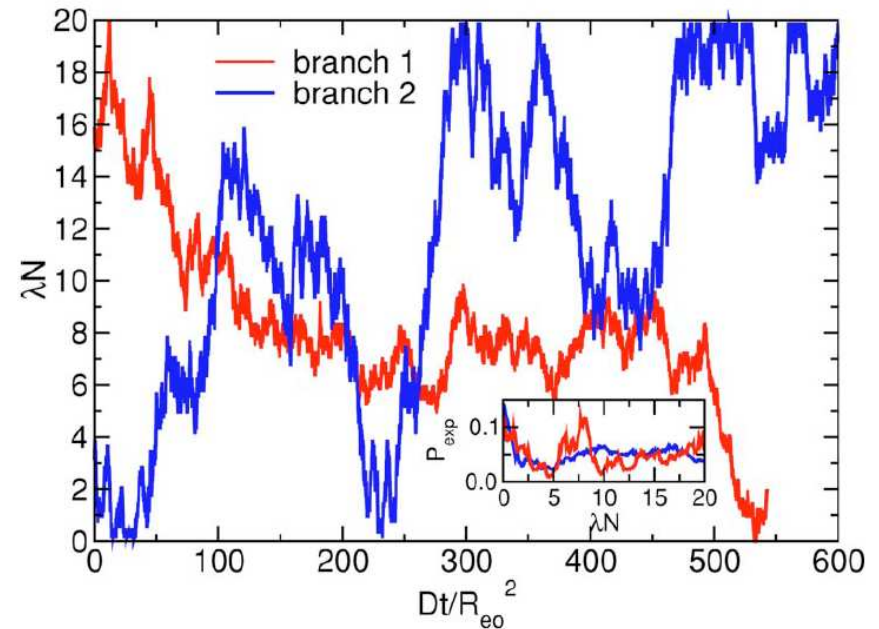
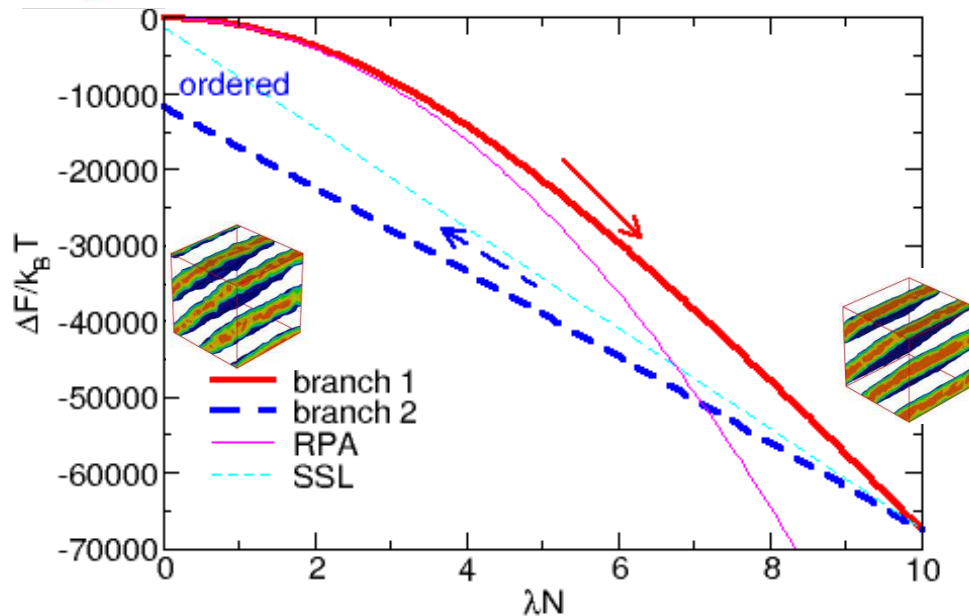
# accuracy of the method

*no kinetic barrier, ie no phase transition*  
*roughly equal probability*



$\Delta F = \mathcal{O}(10^4)k_B T$   
 disordered

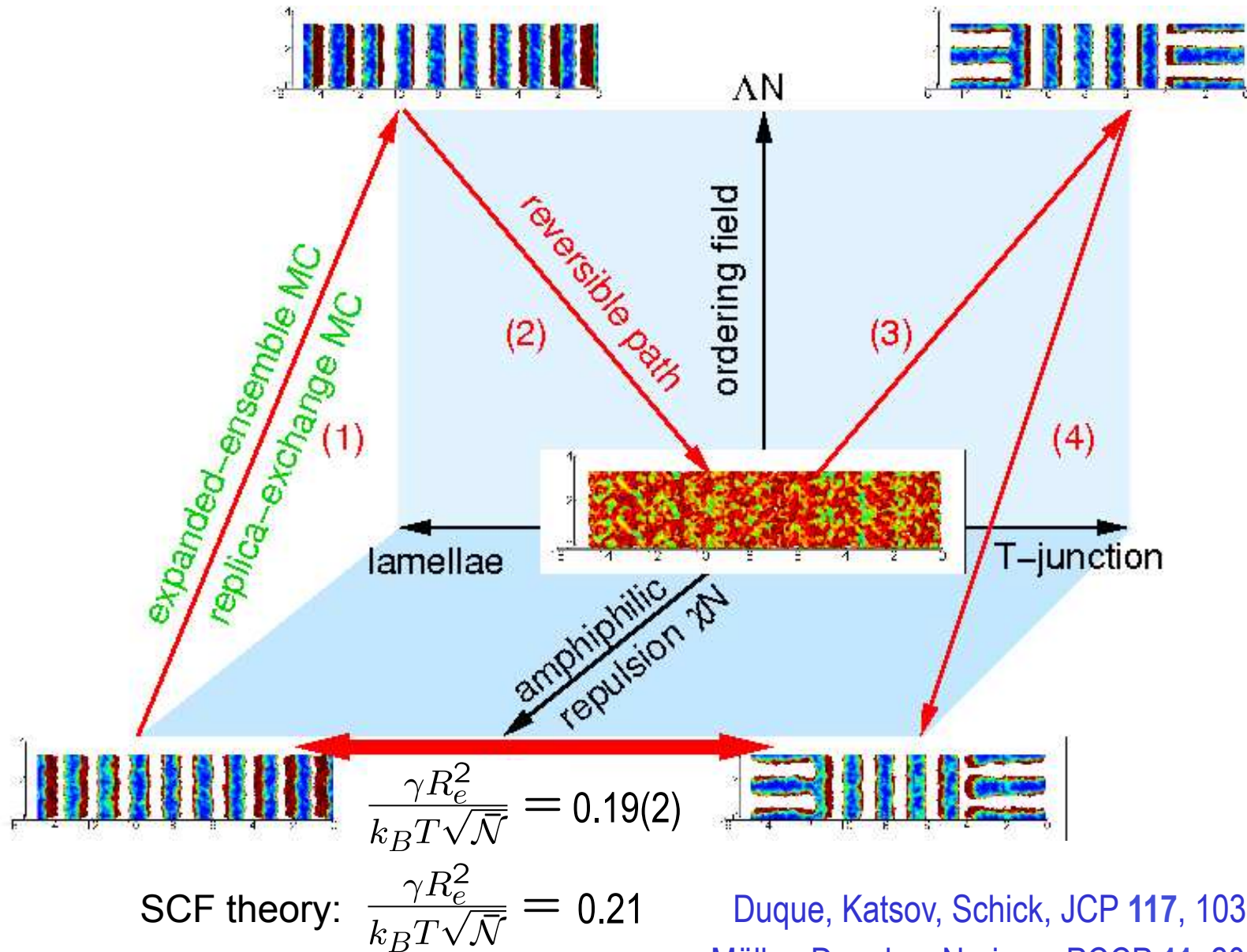
$\delta \Delta F = \mathcal{O}(10^1)k_B T$



$$\Delta F_1 = \int_0^{\lambda N} d\lambda N \left. \frac{\langle \mathcal{H}_{\text{ext}} \rangle}{\lambda N} \right|_{\chi_0 N_{\text{init}}}$$

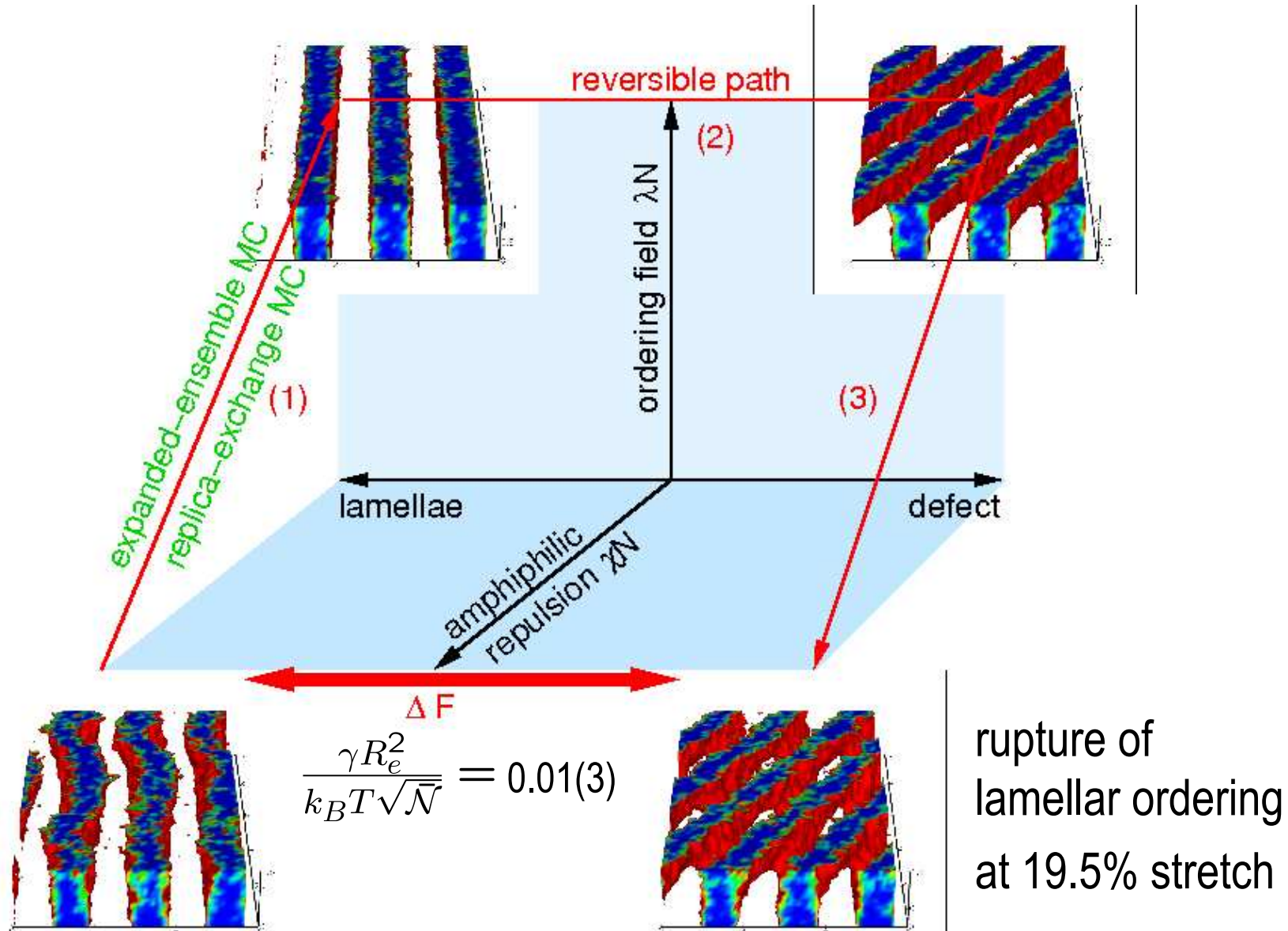
reweighting technique removes large free energy change along the path  
 probability distribution of reweighted simulation estimates accuracy  
 kinetics demonstrates the absence of first-order transition

# free-energy calculations: grain boundaries



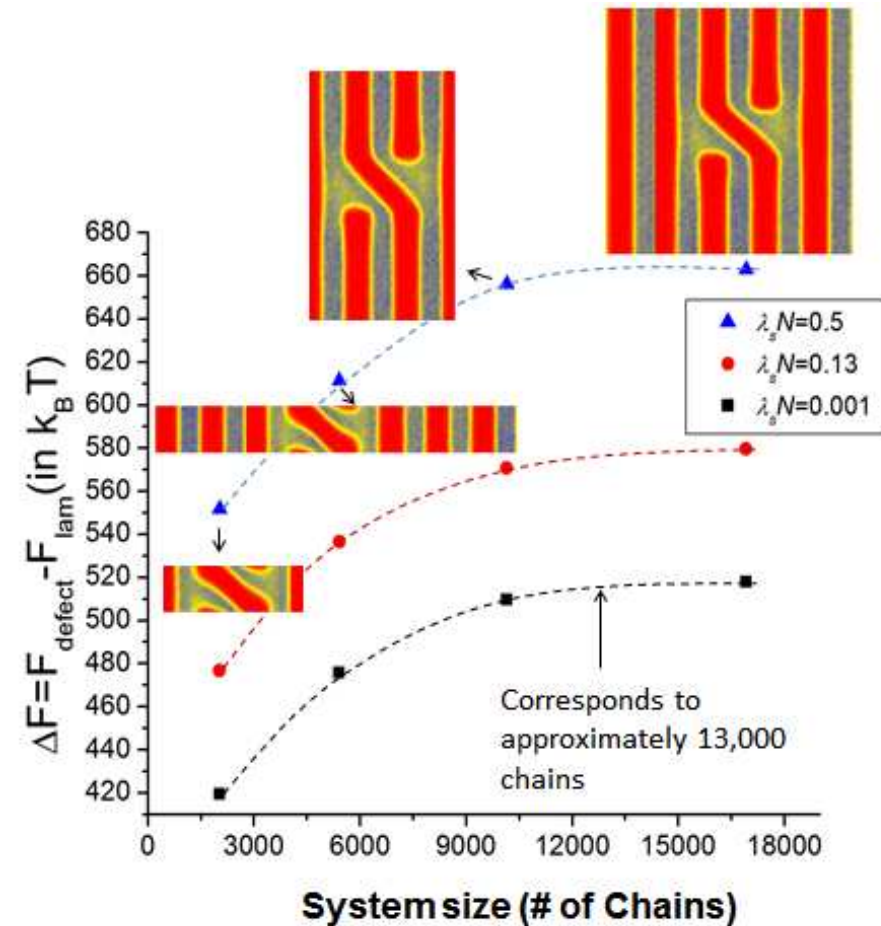
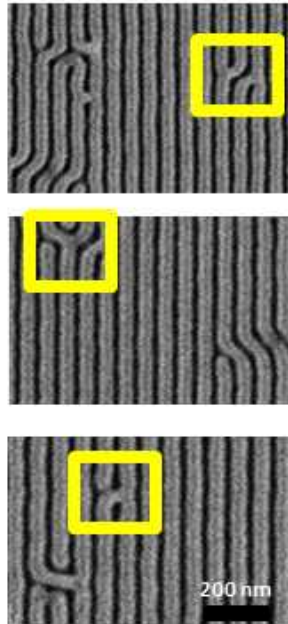
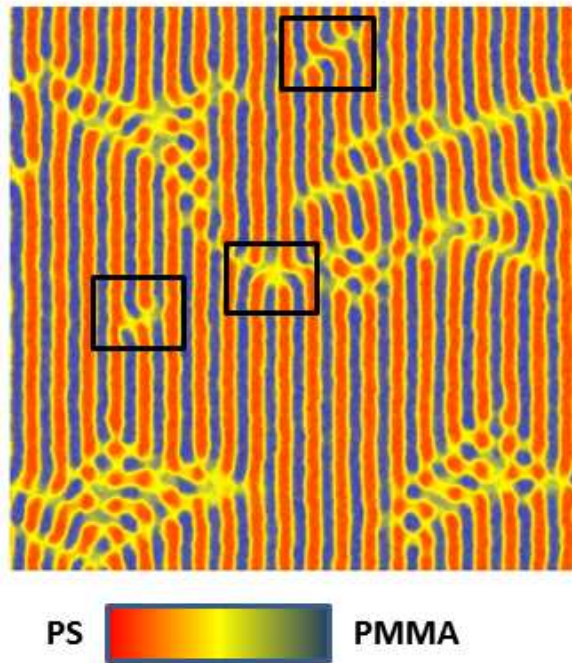
Duque, Katsov, Schick, JCP 117, 10315 (2002)  
Müller, Daoulas, Norizoe, PCCP 11, 2087 (2009)

# reconstruction of soft morphology at patterned surface





# defect free energy for lamellar pattern replication



naïve estimate:

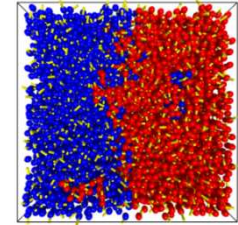
$$\begin{aligned} \Delta F &= \gamma A \\ &\approx \frac{\sqrt{N} k_B T}{R_{eo}^2} \times 2L_s D_o \\ &\approx 128 k_B T \times 3.6 = 460 k_B T \end{aligned}$$

**extremely small defect density in equilibrium**



# particle simulation and continuum description

**system:** symmetric, binary AB homopolymer blend



**degrees of freedom:**

particle coordinates,  $N(n_A + n_B)$   
 $\{\mathbf{r}_i(s)\}$

composition field (and density),  $\infty$

$$m(\mathbf{r}) = \phi_A(\mathbf{r}) - \phi_B(\mathbf{r})$$

$$\rho(\mathbf{r}) = \phi_A(\mathbf{r}) + \phi_B(\mathbf{r}) \approx \rho_0$$

**model definition:**

intra- and intermolecular potentials  
 (here: soft, coarse-grained model, SCMF)

free-energy functional,  $\mathcal{F}_{\text{GL}}[m(\mathbf{r})]$

single-chain dynamics

(Ginzburg-Landau-de Gennes or Ohta-Kawasaki)

(here: Rouse dynamics)

time-dependent GL theory

segmental friction,  $\zeta$

(model B according to Hohenberg & Halperin)

Onsager coefficient,  $\Lambda(\mathbf{r} - \mathbf{r}')$

**projection:**

$$\hat{\phi}_A(\mathbf{r}) \equiv \frac{1}{\rho_0} \sum_{i=1}^{n_A} \sum_{s=1}^N \delta(\mathbf{r} - \mathbf{r}_i(s))$$

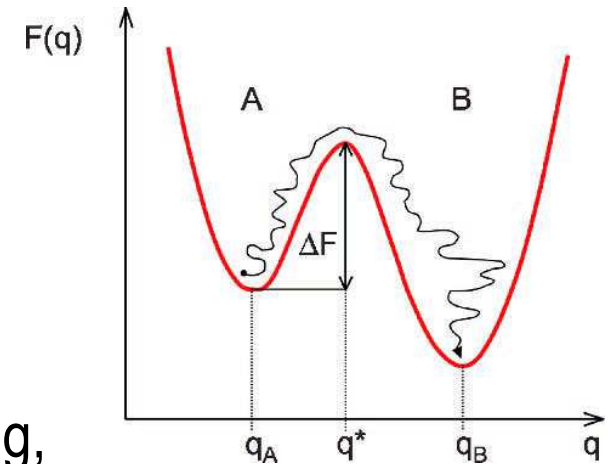
$$\frac{\mathcal{F}[m]}{k_B T} \equiv - \ln \int \mathcal{D}[\{\mathbf{r}_{i,s}\}] e^{-\frac{\mathcal{H}(\{\mathbf{r}_{i,s}\})}{k_B T}} \delta[m - (\hat{\phi}_A - \hat{\phi}_B)]$$

$$\Lambda(\mathbf{r} - \mathbf{r}') = \left\langle \frac{\partial \hat{\phi}(\mathbf{r})}{\partial \mathbf{r}_i(s)} M_{\zeta, i, j}(s, t) \frac{\partial \hat{\phi}(\mathbf{r}')}{\partial \mathbf{r}_j(t)} \right\rangle$$

Kawasaki, Sekimoto, Physica **143A**, 349 (1987)

# speed-up particle simulations by concurrent coupling

*question:* why are particle simulations slow?



## 1) *barrier problem (b):*

system has to overcome a **free-energy barrier**,

Kramer's theory  $\tau \sim \exp(-\Delta F/k_B T)$

solutions: WL sampling, conf.T-WL, conf. flooding,  
metadynamics, transition-path sampling, forward flux sampling, ...

Dellago, Bolhuis, Adv. Polym. Sci **221**, 167 (2008)

## 2) *time-scale problem (a):* “**intrinsically slow processes**”

downhill in continuum free energy but **small Onsager coefficient** (response to TD force) and/or two **vastly different time scales** (stiff equations)

stiff interaction dictates time step, weak interaction drives slow time evolution

solutions: •reversible multiple time step MD (RESPA)

Tuckerman, Berne, Martyna, JCP **97**, 1990 (1992)

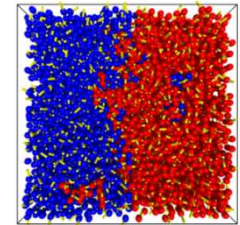
•SCMF simulation

Müller, Smith J.Polym.Sci.B **43**, 934 (2005)

•**HMM**

E, Engquist, Li, Ren, Vanden-Eijnden, Comm. Comp. Phys. **2**, 367 (2007)

# time scale and free-energy separation in polymer blends



**system:** symmetric, binary AB homopolymer blend

bonded interactions (stiff)

$$f_b \sim \frac{k_B T}{b} \sim \frac{k_B T \sqrt{N}}{R_{eo}}$$

non-bonded interactions (weak)

$$f_{nb} \sim \frac{k_B T \chi}{w} \sim \frac{k_B T \chi N}{R_{eo}} N^{-1}$$

➔ **Ginzburg-Landau models do not include stiff bonded interactions and approximate limiting slow time evolution**

**recap: kinetics of phase separation in a symmetric binary polymer blend:**

- spinodal decomposition

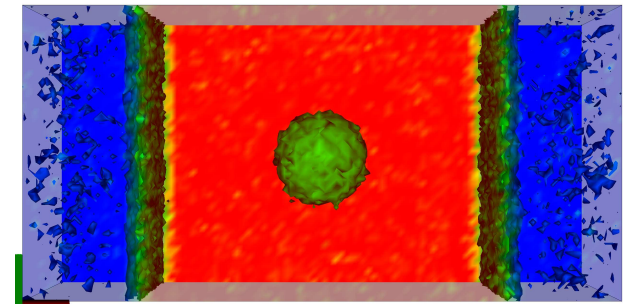
- **Lifshitz-Slyozov coarsening** (diffusive regime)

Gibbs-Thomson equation  $\mu^* = \frac{R_{eo}}{R} \frac{\gamma^*}{m_{coex}}$

$$\frac{\partial m}{\partial t} = \nabla \wedge \nabla \mu$$

$$\frac{1}{t} \sim \frac{1}{L} \wedge \frac{1}{L} \frac{\gamma}{L} \Rightarrow L^3 \sim \gamma \Lambda t$$

$$\frac{d}{dt} \frac{\Delta_o}{N \sqrt{N}} = -\frac{2\pi}{\tau} (1 - m_{coex}^2) \frac{\gamma R_{eo}^2}{k_B T \sqrt{N}}$$



$$\Lambda = \frac{R_{eo}^5}{\tau k_B T \sqrt{N}} (1 - m^2)$$

$$1 - m_{coex} \approx 2 \exp(-\chi_o N)$$

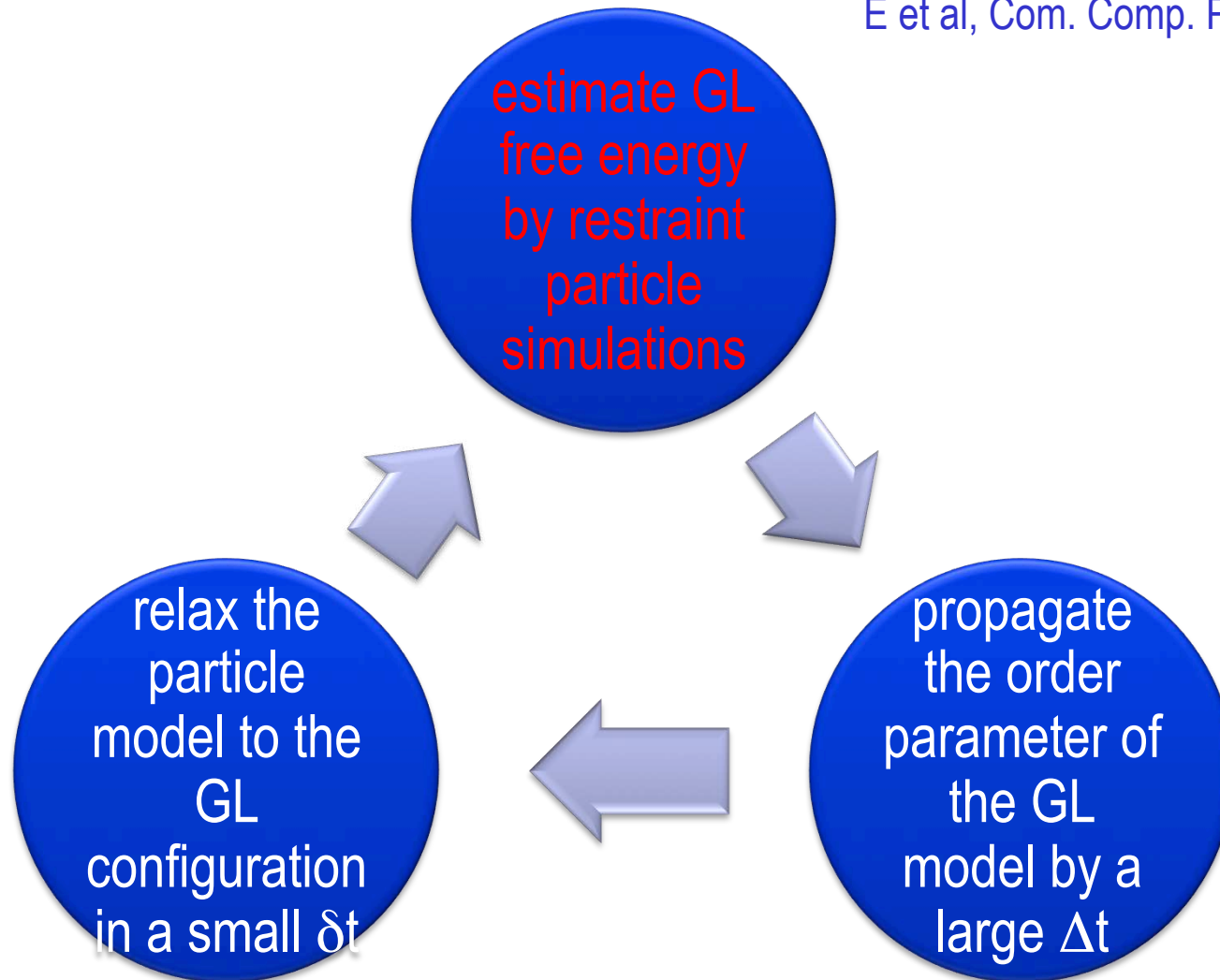
- alternative: droplet coagulation by Brownian motion

irrelevant due to large viscosity of polymer melt

# heterogeneous multiscale modeling (HMM)

E, Ren, Vanden-Eijnden, J. Comp. Phys. **228**, 5437 (2009)

E et al, Com. Comp. Phys **2**, 367 (2007)



to show: steps that involve particle simulation require a time of the order  $\frac{\tau}{\lambda N}$

# free-energy functional from restraint simulations

**idea:** restrain the composition,  $\hat{m} \equiv \hat{\phi}_A - \hat{\phi}_B$ , of particle model to fluctuate around the order-parameter field,  $m(\mathbf{r})$ , of the continuum description (**field-theoretic umbrella sampling** for order-parameter field,  $m(\mathbf{r})$ )

$$\frac{\mathcal{H}_b[\mathbf{r}_i(s)]}{k_B T} = \sum_{s=1}^{N-1} \frac{3(N-1)}{2R_{eo}^2} [\mathbf{r}_i(s) - \mathbf{r}_i(s+1)]^2$$

$$\frac{\mathcal{H}_{nb}[\hat{\phi}_A, \hat{\phi}_B]}{k_B T \sqrt{\mathcal{N}}} = \int \frac{d^3 \mathbf{r}}{R_{eo}^3} \left( \frac{\kappa_o N}{2} [\hat{\phi}_A + \hat{\phi}_B - 1]^2 - \frac{\chi_o N}{4} [\hat{\phi}_A - \hat{\phi}_B]^2 \right)$$

$$\frac{\mathcal{H}_{\lambda N}}{k_B T \sqrt{\mathcal{N}}} = \frac{\lambda N}{2} \int \frac{d^3 \mathbf{r}}{R_{eo}^3} \left\{ \left[ \hat{\phi}_A - \frac{1+m}{2} \right]^2 + \left[ \hat{\phi}_B - \frac{1-m}{2} \right]^2 \right\}$$

$\lambda N \gg \chi_o N$  strong coupling between particle model and continuum description

$$\exp \left( -\frac{\mathcal{H}_{\lambda N}}{k_B T} \right) \xrightarrow{\lambda N \rightarrow \infty} \delta \left( m(\mathbf{r}) - \hat{\phi}_A + \hat{\phi}_B \right) \delta \left( \hat{\phi}_A + \hat{\phi}_B - 1 \right)$$

$$\mu(\mathbf{r}) = \frac{\delta \mathcal{F}}{\delta m(\mathbf{r})} \xrightarrow{\lambda N \rightarrow \infty} \frac{\delta \mathcal{F}_{\lambda N}}{\delta m(\mathbf{r})} = \left\langle \frac{\delta \mathcal{H}_{\lambda N}}{\delta m(\mathbf{r})} \right\rangle$$

$$\mu^* \equiv \frac{\mu R_{eo}^3}{k_B T \sqrt{\mathcal{N}}} \xrightarrow{\lambda N \rightarrow \infty} \frac{\lambda N}{2} \left( m(\mathbf{r}) - \langle \hat{\phi}_A(\mathbf{r}) - \hat{\phi}_B(\mathbf{r}) \rangle \right) = \frac{\lambda N}{2} \left( m(\mathbf{r}) - \langle \hat{m}(\mathbf{r}) \rangle \right)$$

inspired by Maragliano, Vanden-Eijnden, Chem. Phys. Lett. **426**, 168 (2006)

# free-energy functional from restraint simulations

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(**field-theoretic umbrella sampling** for order-parameter field,  $m(\mathbf{r})$ )

make an **Ansatz for the continuum description** with a few parameters  $\{a_r\}, a_g$   
determine parameters by comparing  $\mu(\mathbf{r}|m)$  with result of Ansatz

GL *model* required, not “equation-free” Kevrekidis, Gear, Hummer *AICHE J.* **50**, 1346 (2004)  
**average over space** (instead of time) to determine the few parameters  
of the Ginzburg-Landau model (spatial homogeneity of GL model)

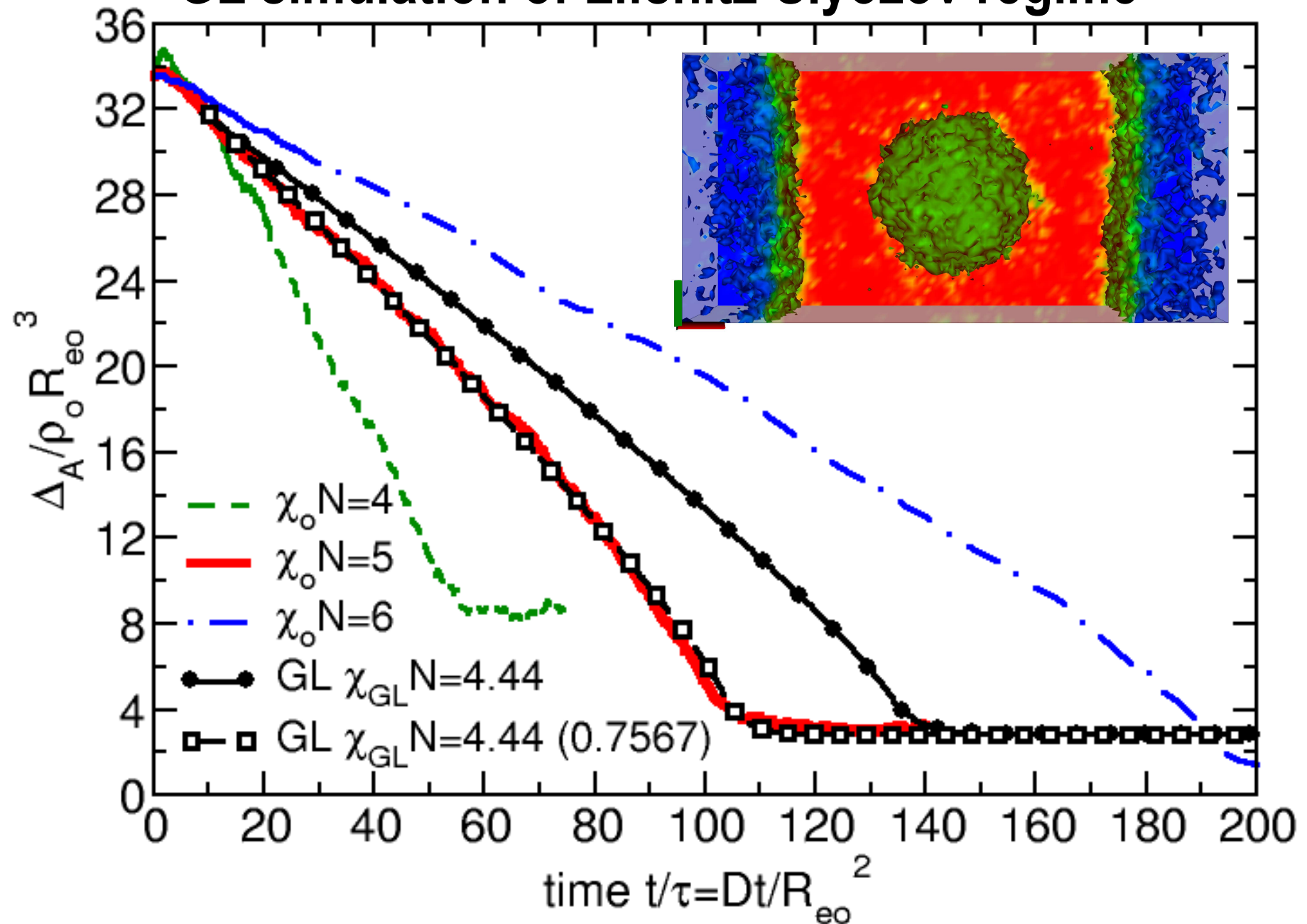
result: Ginzburg-Landau model for a particle model at a specific state

**large time step  $\Delta t$  is limited by the condition that parameters do not vary on scale  $\Delta t$**

e.g., intrinsic structure of interface must not vary WSL vs SSL  
but location of interface may move



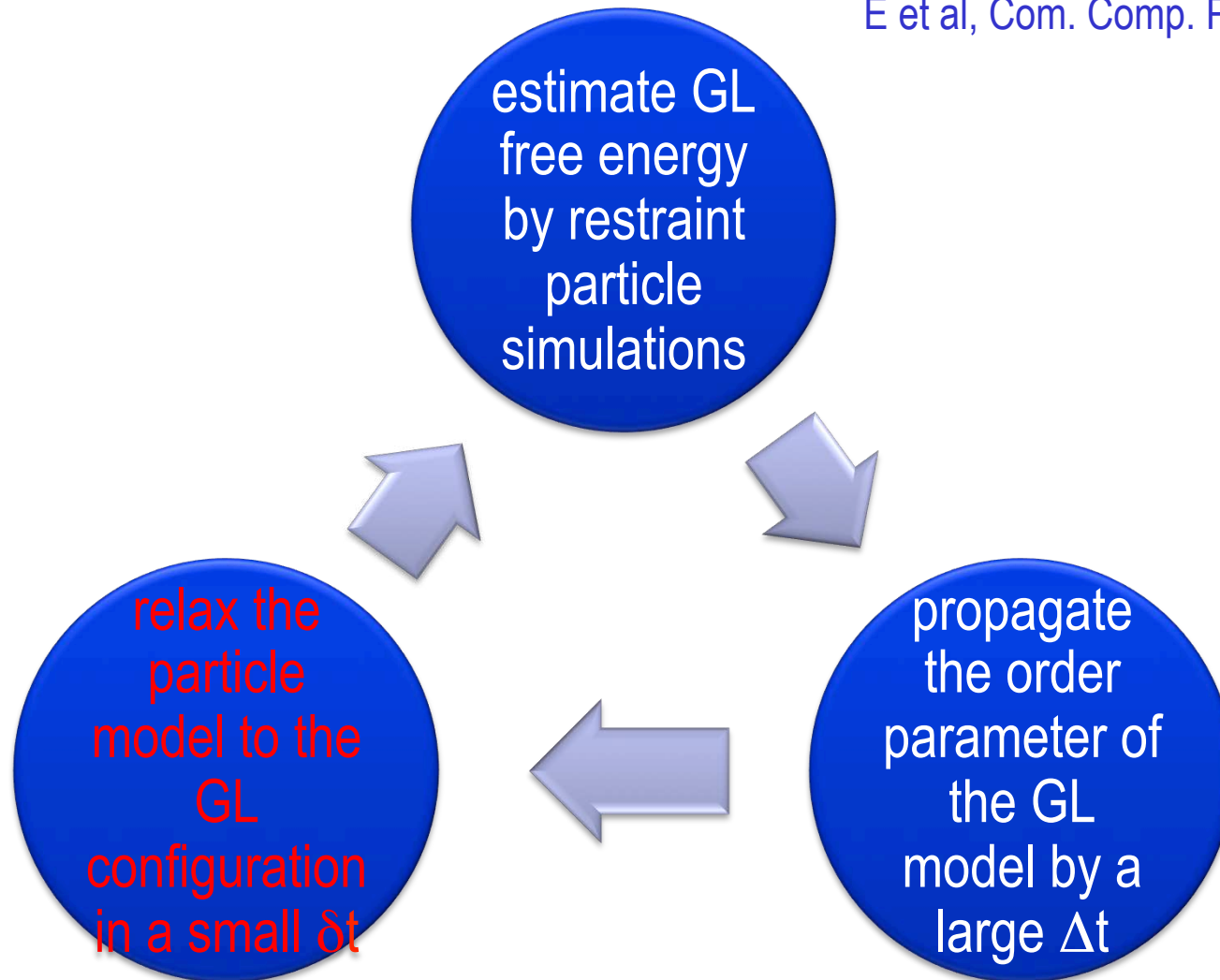
# GL simulation of Lifshitz-Slyozov regime



# heterogeneous multiscale modeling (HMM)

E, Ren, Vanden-Eijnden, J. Comp. Phys. **228**, 5437 (2009)

E et al, Com. Comp. Phys **2**, 367 (2007)



to show: steps that involve particle simulation require a time of the order  $\frac{\tau}{\lambda N}$

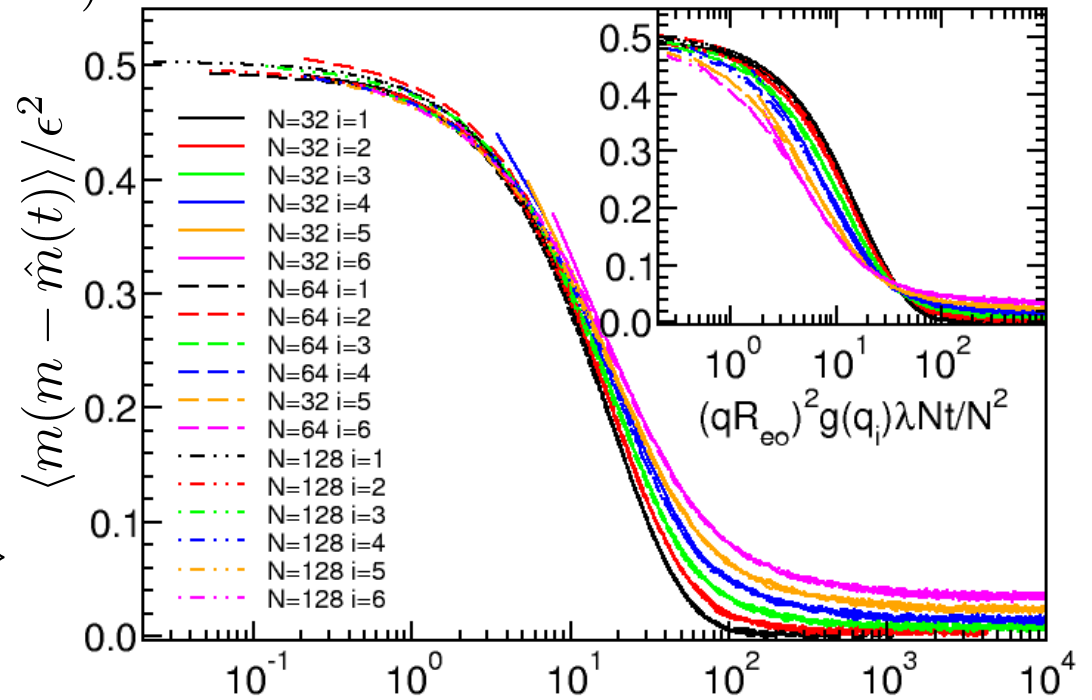
# Onsager coefficient from relaxation of simulations

**idea:** study relaxation of restraint system towards equilibrium,  $\langle \hat{m} \rangle$

relaxation time of the constraint system is speeded-up by a factor  $\lambda N$

$$\frac{\mathcal{F}_{\lambda N}[\hat{m}]}{k_B T \sqrt{N}} = \frac{\mathcal{F}[\hat{m}]}{k_B T \sqrt{N}} + \frac{\lambda N}{4} \int \frac{d^3 \mathbf{r}}{R_e^3} (m - \hat{m})^2 \quad L = 6R_{e0}, \epsilon = 0.5, \lambda N = 100, q_i = \frac{2\pi i}{L}$$

$$\mu_{\lambda N}^*(\mathbf{r}) = \mu^*[\hat{m}] + \frac{\lambda N}{2} (\hat{m} - m)$$



$$\tilde{\mu}_{\lambda N}^*(\mathbf{q}) = \frac{1}{2} \left\{ \frac{\hat{m}}{2h(\mathbf{q})} + \lambda N (\hat{m} - \tilde{m}) \right\}$$

$$\tilde{\Lambda}(\mathbf{q}) \approx \frac{R_e^5}{V k_B T \sqrt{N}} \frac{g(\mathbf{q})}{\tau_{GL}}$$

$$\frac{\partial \hat{m}(\mathbf{q}, t)}{\partial t} = - \frac{(qR_{e0})^2 g(\mathbf{q})}{2\tau_{GL}} \left\{ \lambda N + \frac{1}{2h(\mathbf{q})} \right\} (\hat{m} - \langle \hat{m} \rangle) \quad (qR_{e0})^2 \lambda N t / N^2 \sim (qR_{e0})^2 \lambda N \frac{t}{\tau}$$

➔ constraint system exponentially relaxes towards  $\langle \hat{m} \rangle = \frac{\tilde{m}}{1 + \frac{1}{2\lambda N h(\mathbf{q})}}$   
 with a **fast** relaxation time scale  $\tau_{\lambda N} = \frac{2\tau_{GL}}{\lambda N} \frac{1}{(qR_e)^2 g(\mathbf{q})}$  (fraction of Rouse time)

# speed-up and scale separation

**question:** What limits the increase of  $\lambda N$  ?

accurate measurement of the chemical potential  $\lambda N \sim \frac{V\sqrt{N}}{R_{eo}^3}$   
 forces due to the restraint must be smaller than the original forces  
 that dictate the intrinsic kinetics of the particle model

bonded force per segment

non-bonded, thermodynamic force

restraint force

$$f_b \sim \frac{k_B T}{b} \sim \frac{k_B T \sqrt{N}}{R_{eo}}$$

$$f_{nb} \sim \frac{k_B T \chi}{w} \sim \frac{k_B T \chi N}{R_{eo}} N^{-1}$$

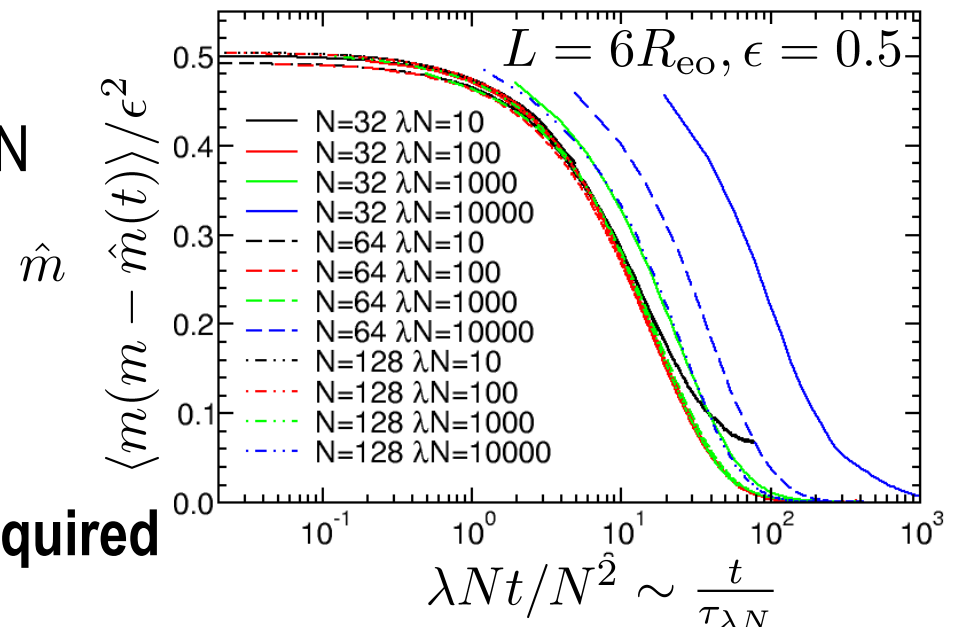
$$f_{\lambda N} \sim \frac{k_B T \lambda}{\Delta L} \sim \frac{k_B T \lambda N}{R_{eo}} N^{-1/2}$$

$$f_{nb} \ll f_{\lambda N} \ll f_b \implies \chi_o N \ll \lambda N \sqrt{N} \ll N^{3/2}$$

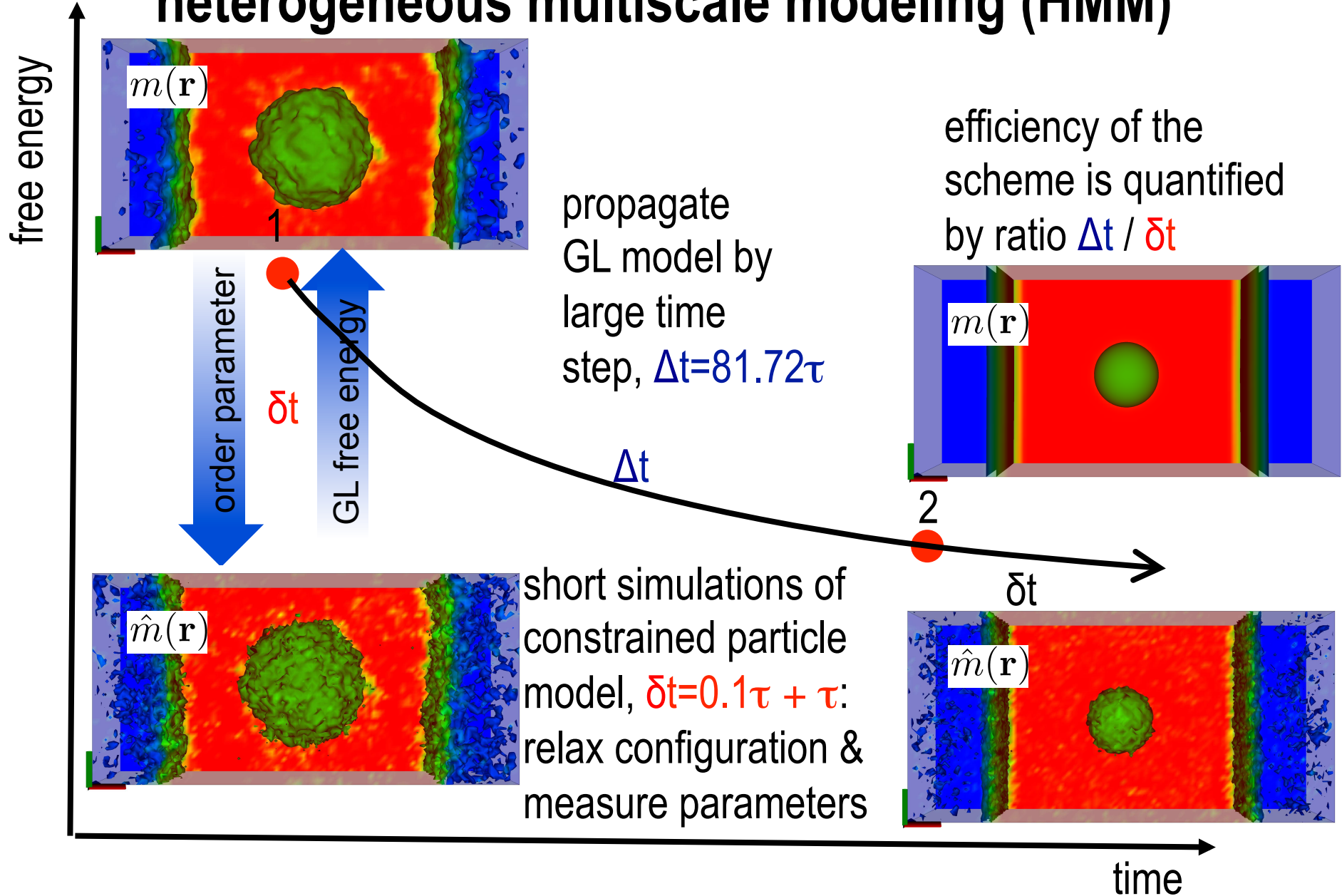
relaxation rate is increase by a factor  $\lambda N$   
 for small composition/density variations

**caveat:**  $\lambda N \geq 1$  not linear response,  
 dynamic RPA fails

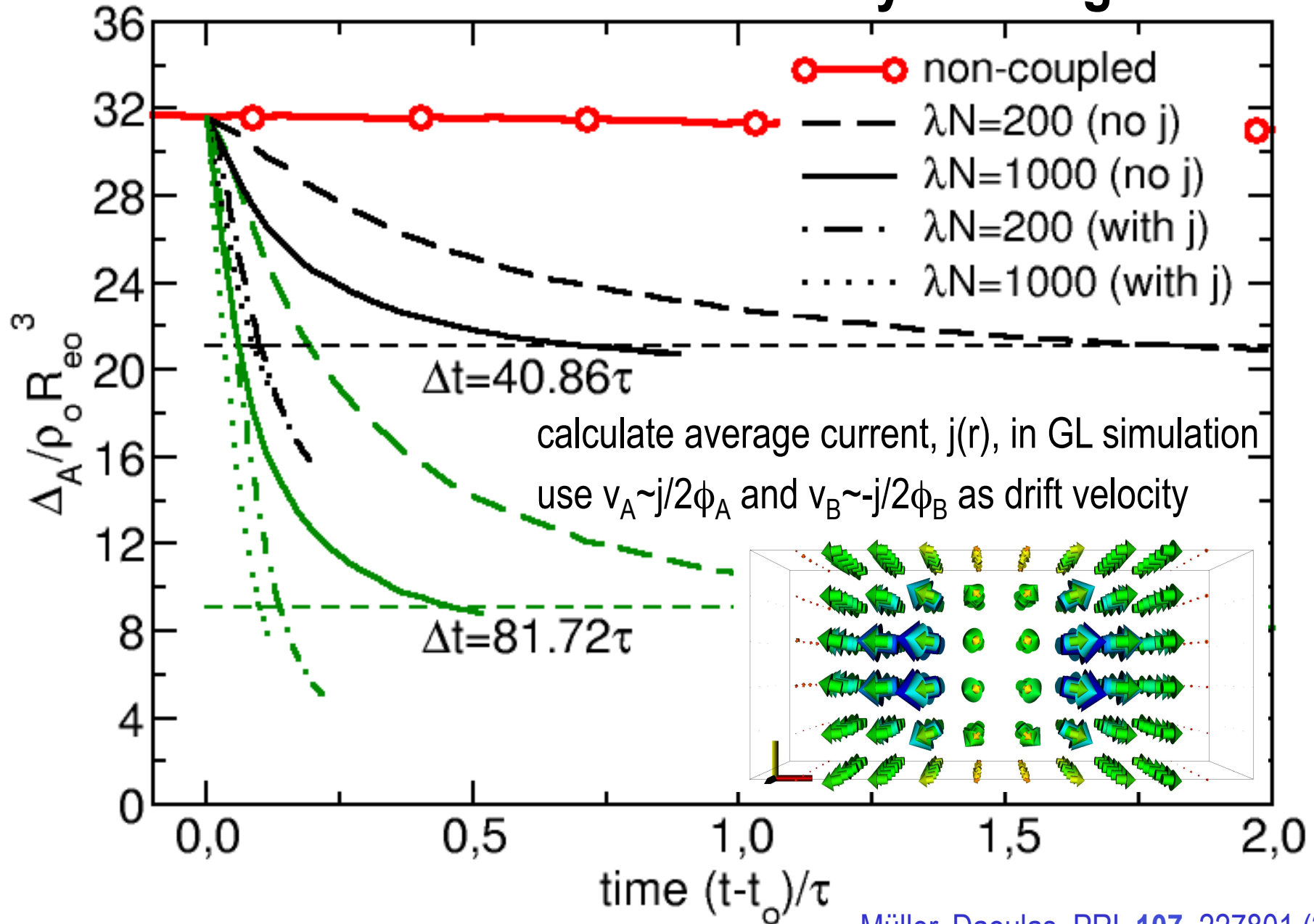
**additional relaxation  $\tau$  may be required**



# heterogeneous multiscale modeling (HMM)

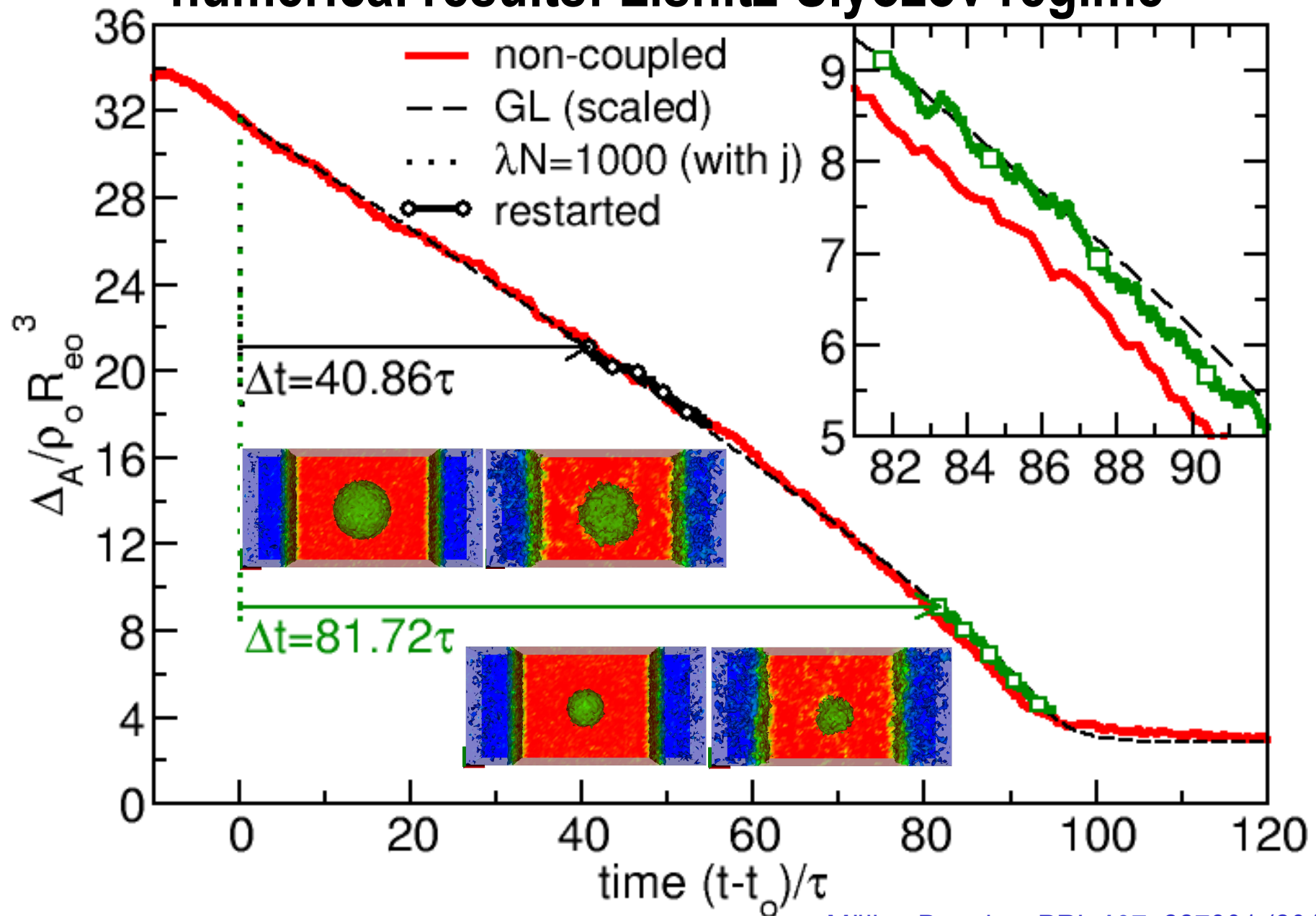


# numerical results: Lishitz-Slyozov regime





# numerical results: Lishitz-Slyozov regime



## b) collective transformation between morphologies

**problem:** find a **physical reversible path**  
that identifies the barrier to stalk formation

which “coordinates” describe transformation?  
particle coordinates in liquid are impractical because  
(i) missing entropy and (ii) permutation symmetry

➔ use order parameter field  $m(\mathbf{r}) = \phi_A(\mathbf{r}) - \phi_B(\mathbf{r})$   
DFT suggests that collective densities are suitable

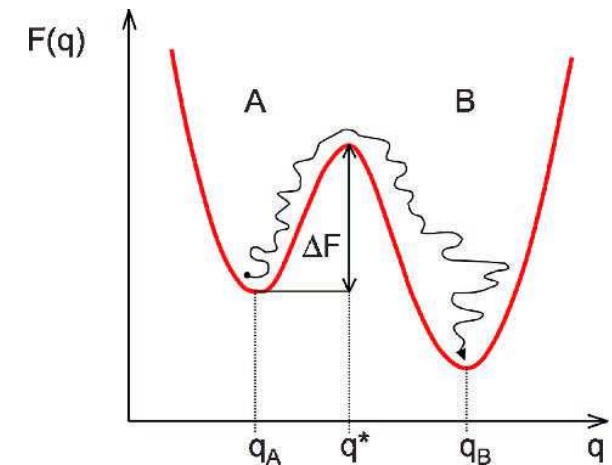
compute free-energy functional

$$\frac{\mathcal{F}[m]}{k_B T} = -\ln \int \mathcal{D}[\{\mathbf{r}\}] e^{-\frac{\mathcal{H}[\{\mathbf{r}\}]}{k_B T}} \delta[m(\mathbf{r}) - \hat{m}(\mathbf{r})]$$

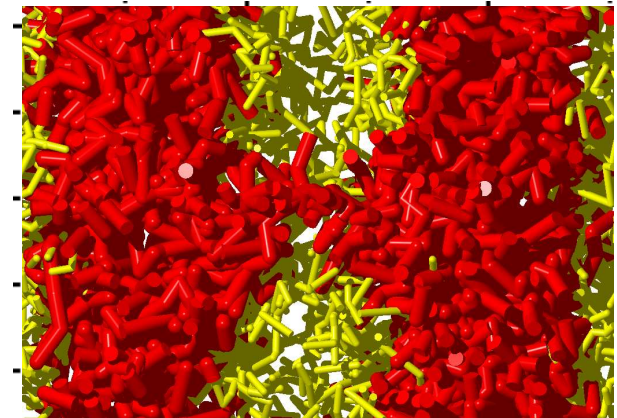
➔ field-theoretic umbrella sampling,  
on-the-fly string method (particle-based)

find minimal free-energy path (MFP) in high-  
dimensional “coordinate” space (for functional  $\mathcal{F}[m]$  on collocation lattice)

➔ improved string method



Dellago, Bolhuis, *Adv. Polym. Sci.* **221**, 167 (2008)



E, Ren, Vanden-Eijnden, *JCP* **126**, 164103 (2007)  
Maragliano, Vanden-Eijnden. *Chem. Phys. Lett.*, **446**, 182 (2007)

# on-the-fly string method and improved string method

describe the transformation path by a string of morphologies  $m_s(\mathbf{r})$  that is parameterized by a contour variable  $0 \leq s \leq 1$

the minimum free-energy path (MFP) is defined by condition that the derivative perpendicular to the path vanishes

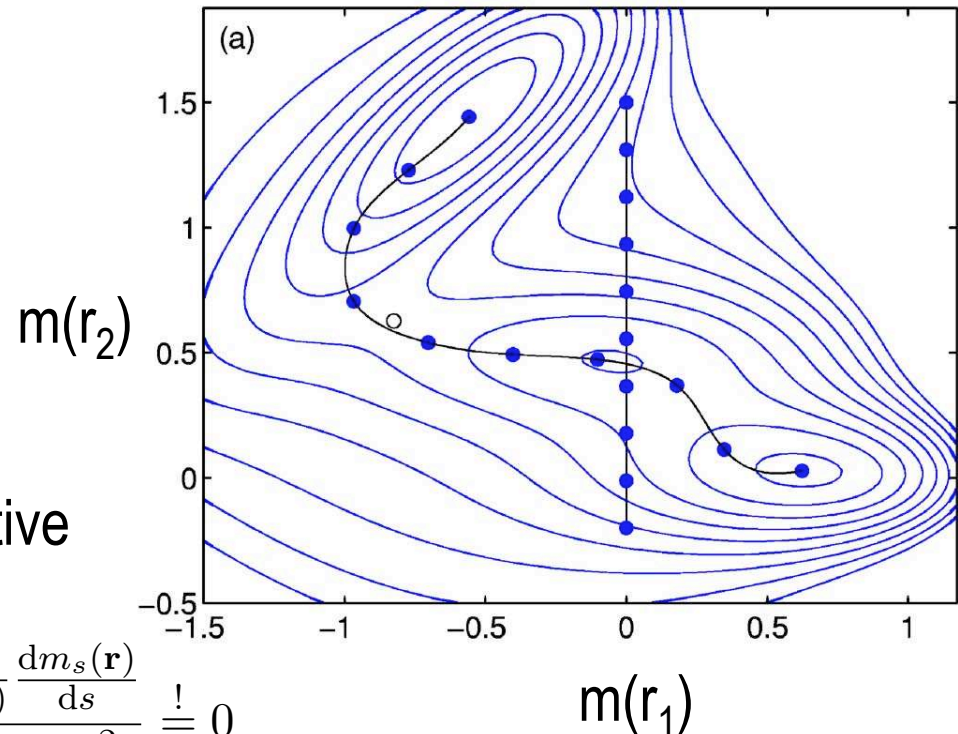
$$\nabla_{\perp} \mathcal{F}[m] = \frac{\delta \mathcal{F}}{\delta m_s(\mathbf{r})} - \frac{dm_s(\mathbf{r})}{ds} \frac{\int d^3\mathbf{r} \frac{\delta \mathcal{F}}{\delta m_s(\mathbf{r})} \frac{dm_s(\mathbf{r})}{ds}}{\int d^3\mathbf{r} \left( \frac{dm_s(\mathbf{r})}{ds} \right)^2} \stackrel{!}{=} 0$$

on-the-fly string method and improved string method:

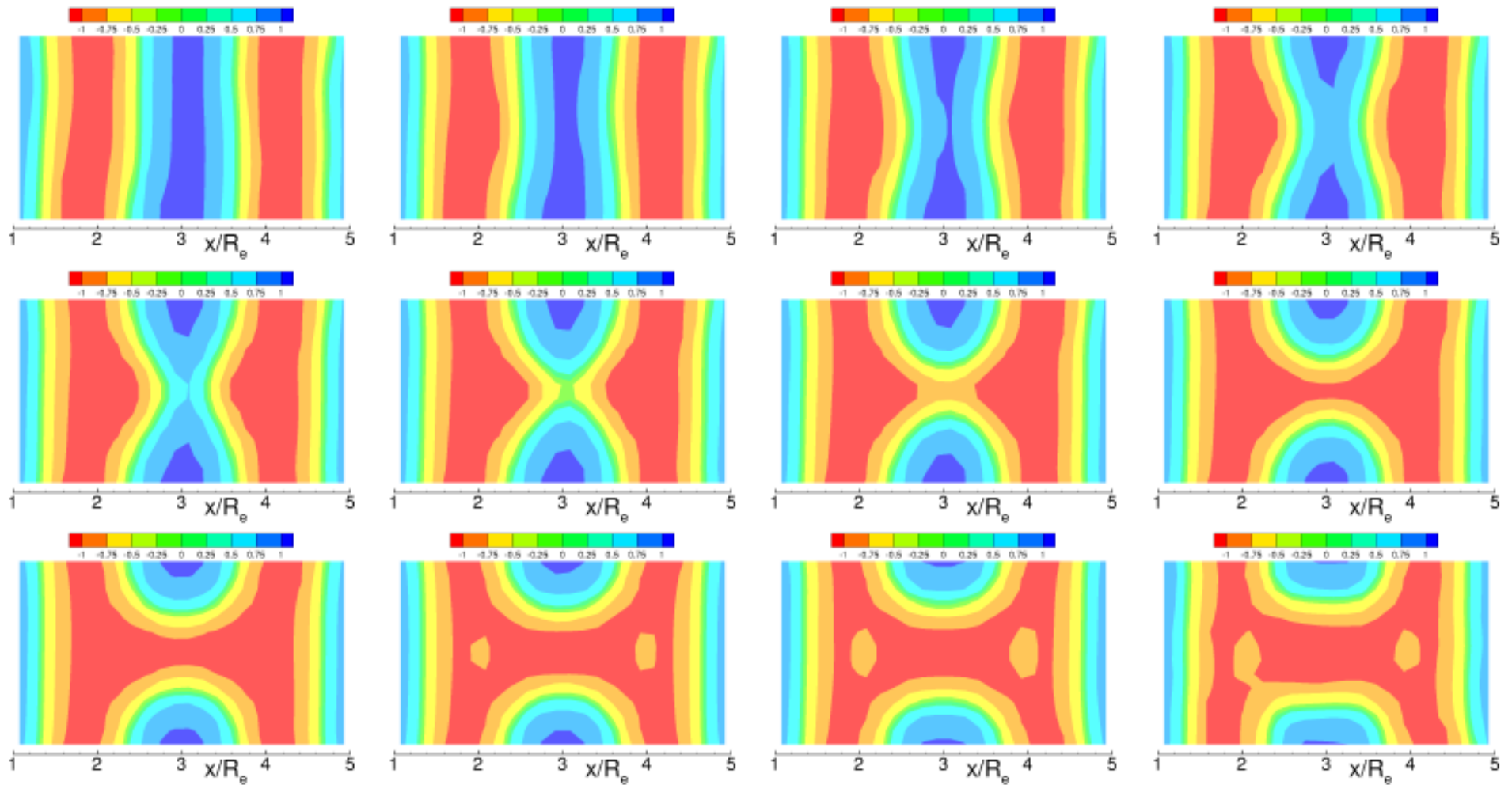
evolve each morphology  $m_s(\mathbf{r})$  as to minimize the free energy

$$\frac{\delta F_c[m_c]}{\delta m_c(\mathbf{r})} = \lambda k_B T [m_c(\mathbf{r}) - \langle \hat{m}(\mathbf{r}) \rangle_c] \xrightarrow{\lambda \rightarrow \infty} \mu(\mathbf{r}|m_c) \quad \Delta m_s(\mathbf{r}) = -\mu(\mathbf{r}|m_s) \Delta$$

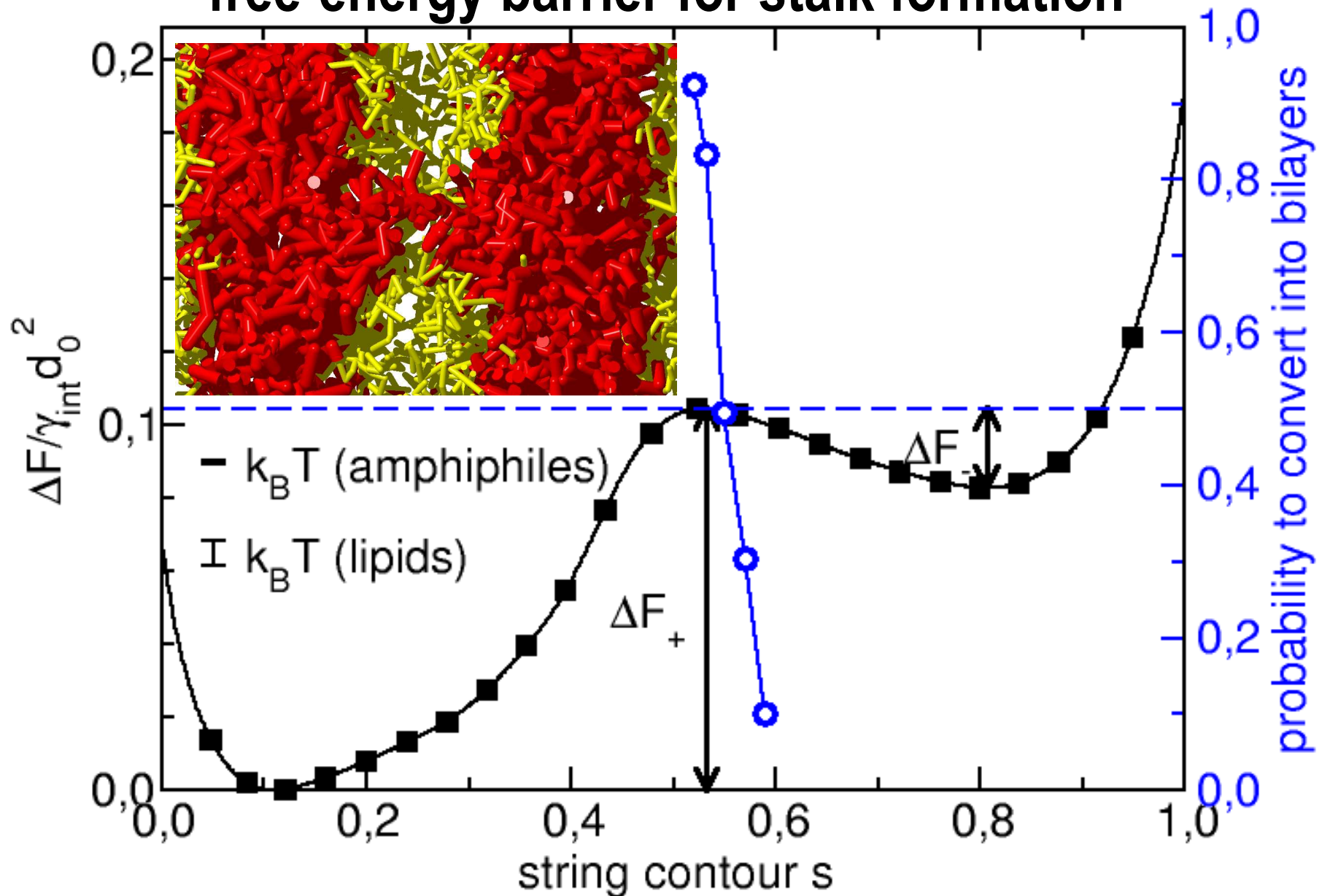
re-parameterize the string to equal distance  $\Delta s$  (pointwise 3<sup>rd</sup> order spline)



# string of morphologies for stalk formation

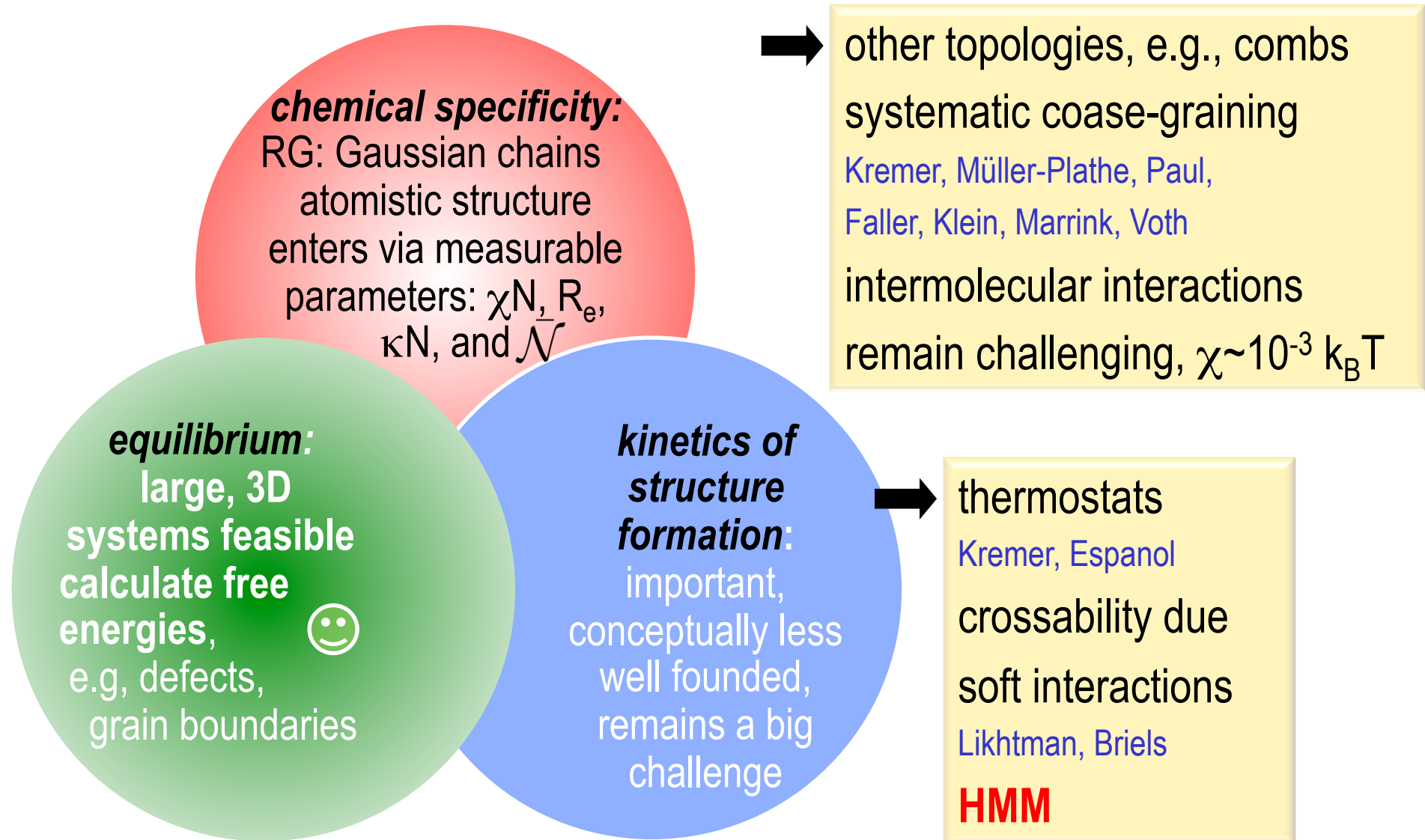


# free-energy barrier for stalk formation





# Mesososcopic and Monte Carlo Simulations



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**Juan de Pablo / Paul Nealey**

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