

Mesoscopic and Monte Carlo Simulations



Marcus Müller

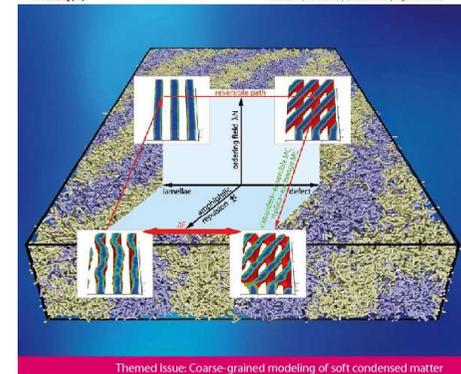
- recap of basic concepts (Mikhail Tamm, Alexey Shaytan): thermodynamics, statistical mechanics, phase transitions
- Monte-Carlo simulations: detailed balance, force-bias Monte-Carlo, phase coexistence
- free-energy of soft-condensed matter: thermodynamic integration in a spatially varying field
- speeding-up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

PCCP

Physical Chemistry Chemical Physics

www.rsc.org/pccp

Volume 12 | Number 12 | 28 March 2010 | Pages 1853-2128



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

COMMUNICATION
Bauer and Fleber
Dispersive particle dynamics: simulation
of quaternary block copolymers



Published by
American Physical Society.



Volume 108, Number 22



GEORG-AUGUST-UNIVERSITÄT
GÖTTINGEN

Moscow, July 18, 2012

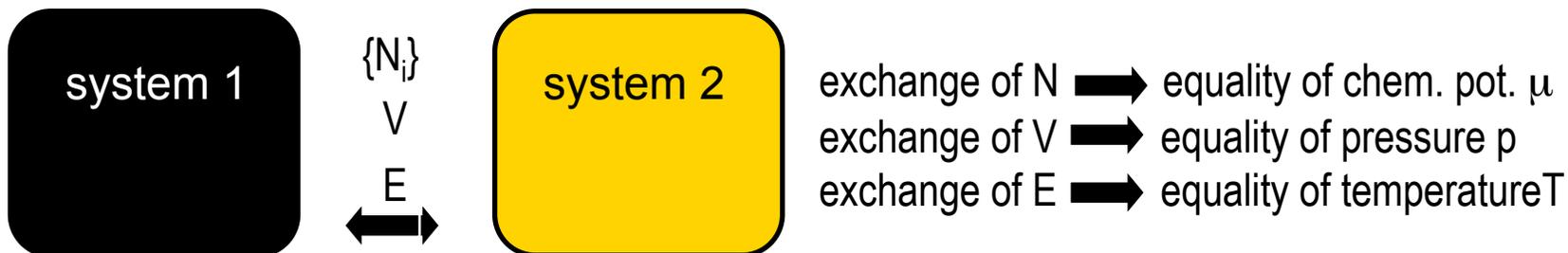
basics: thermodynamics

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 Γ on energy hypersurface, $\mathcal{H}(x_i, p_i) = E$

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

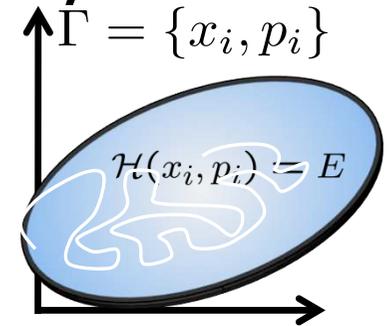
$$\begin{aligned} \bar{A}(\Gamma_0) &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T dt A(\Gamma(t, \Gamma_0)) \\ &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T dt \int_{\Gamma} d\Gamma' A(\Gamma') \delta[\Gamma' - \Gamma(t, \Gamma_0)] \\ &= \int_{\Gamma} d\Gamma' A(\Gamma') \lim_{T \rightarrow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

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**

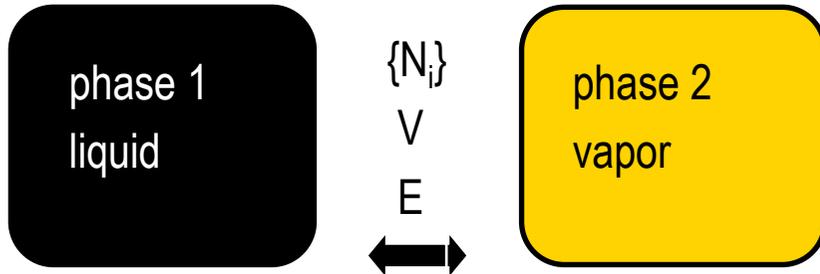
$$S[\rho] = -k_B \frac{1}{N! h^{3N}} \int d\Gamma \rho(\Gamma) \ln \rho(\Gamma) \longrightarrow S(N, V, E) = -k_B \langle \ln \rho_{\text{mc}}(\Gamma) \rangle = k_B \ln \mathcal{Z}_{\text{mc}}$$

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: 1st derivative of thermodyn. potential (state variable)

is not continuous

➡ miscibility gap

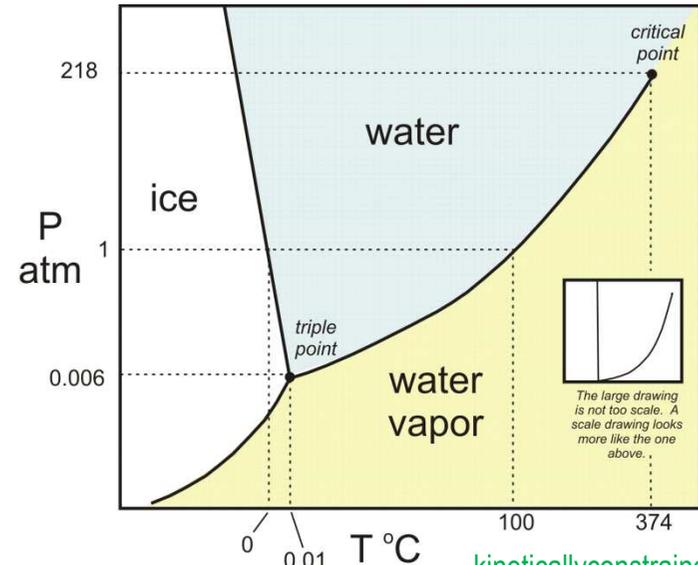
$$N = - \left. \frac{\partial \Omega}{\partial \mu} \right|_{V, T}$$

second order: 2nd derivative of TP (response function)

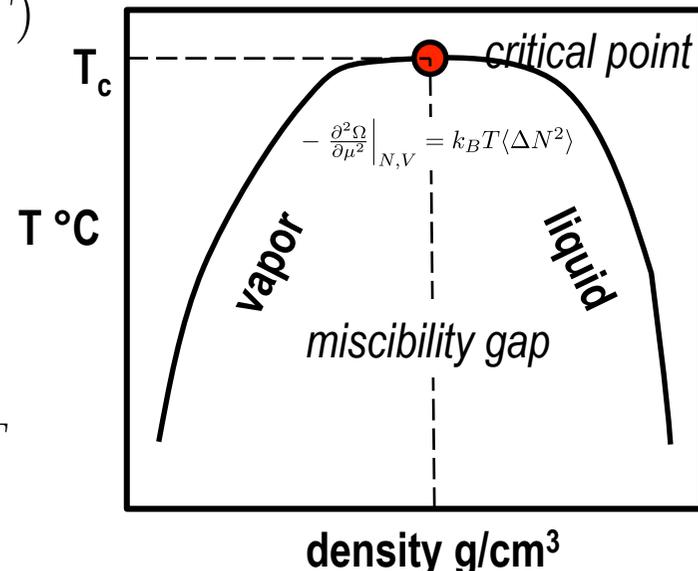
is not continuous

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

➡ compressibility diverges at critical point



kineticallyconstrained.blogspot.com/2009/05/critical-point.html



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 ρ

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

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

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

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

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



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

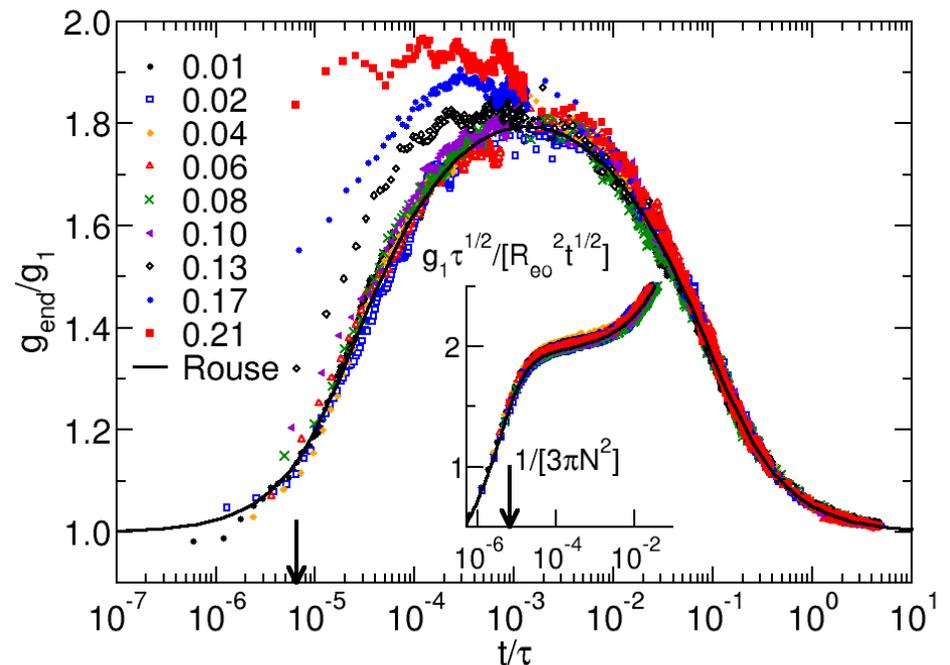
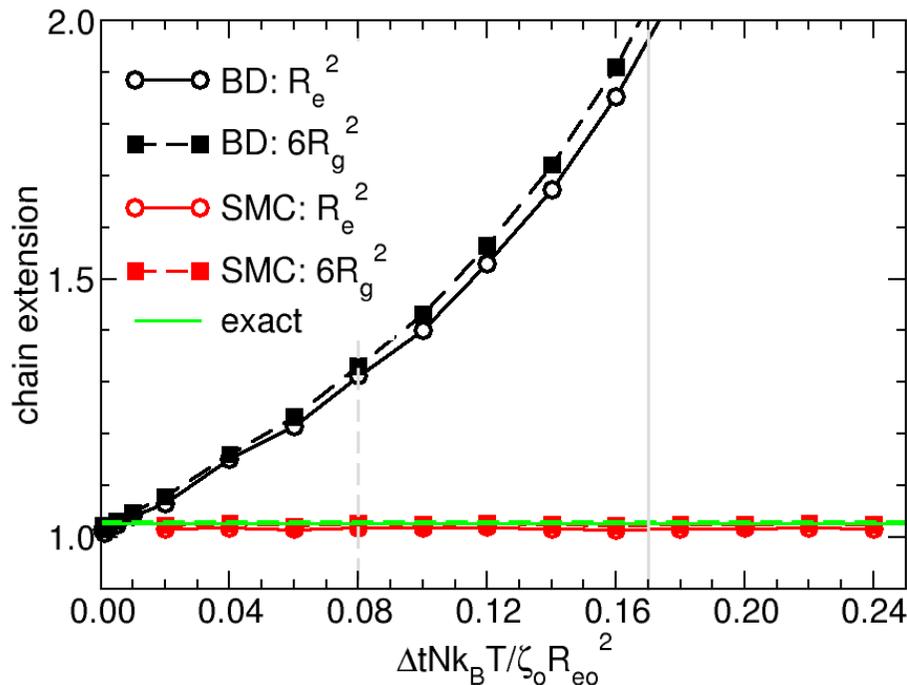
Rouse-like dynamics via Smart-MC simulations

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

idea: uses forces to construct trial displacements Δr

application: soft, coarse-grained polymer model

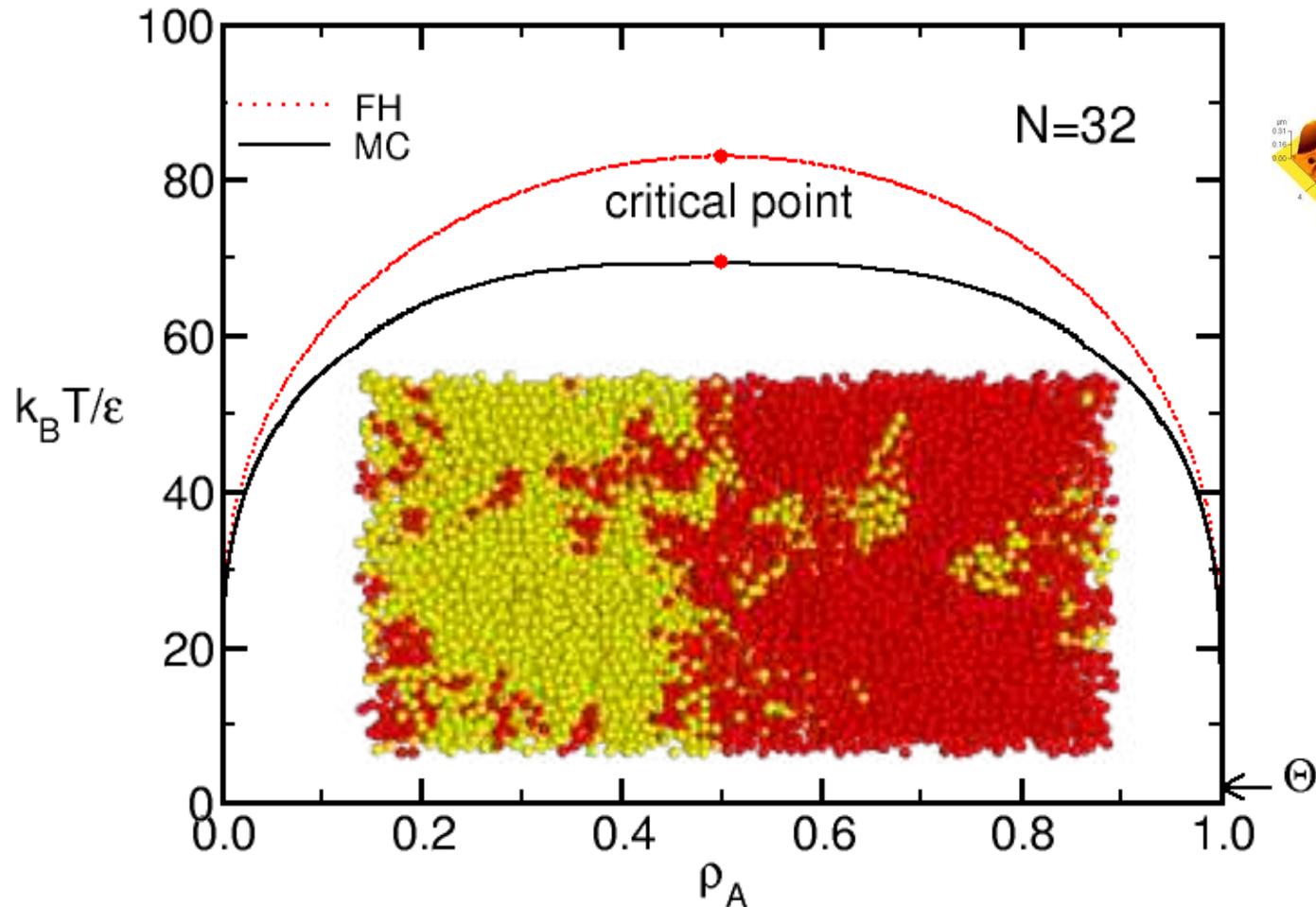
- SMC or force-bias MC allows for a larger time step (factor 100) than BD with Euler integration
- rather realistic dynamics (diffusive) after a few Smart-Monte-Carlo steps



prototypical example: liquid-liquid phase separation in blends

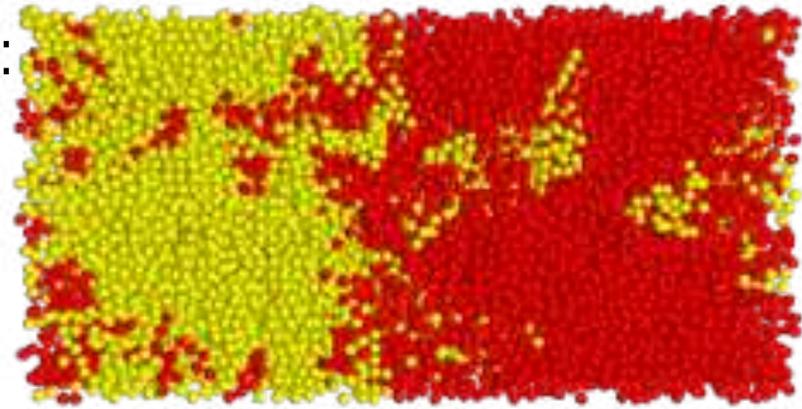
demixing into A-rich and B-rich regions, interfaces in a polymer blend

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:

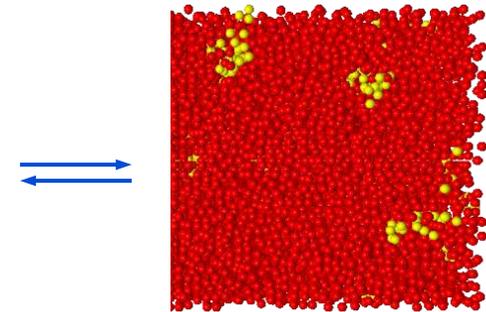
requires very large systems to distinguish interface and bulk properties

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:

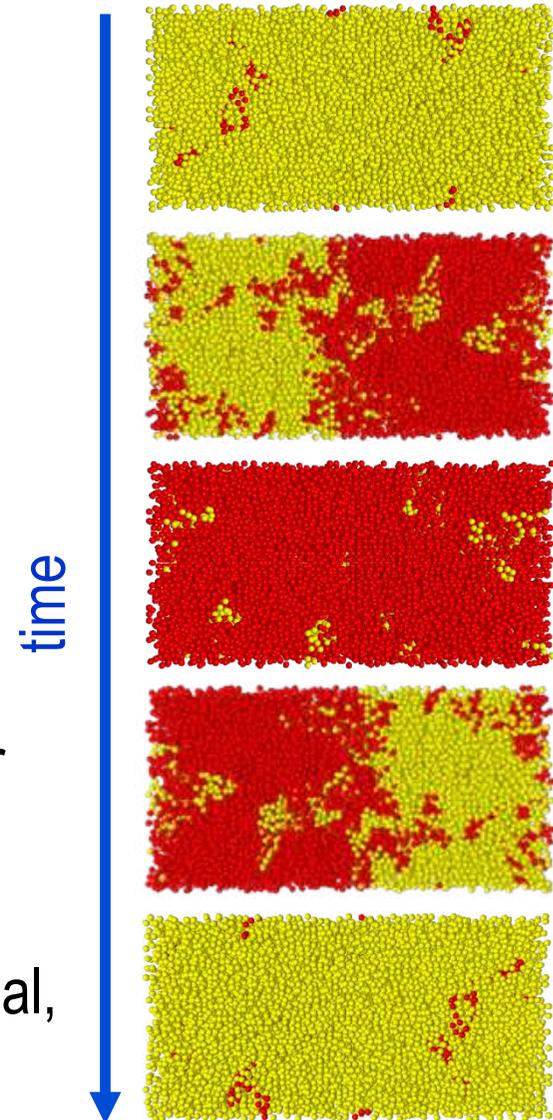
control temperature T , volume V , and total number of polymers, exchange potential, $\Delta\mu$ (semi-gc) or chemical potential, μ (grandcanonical)
order parameter fluctuates

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 μ 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:

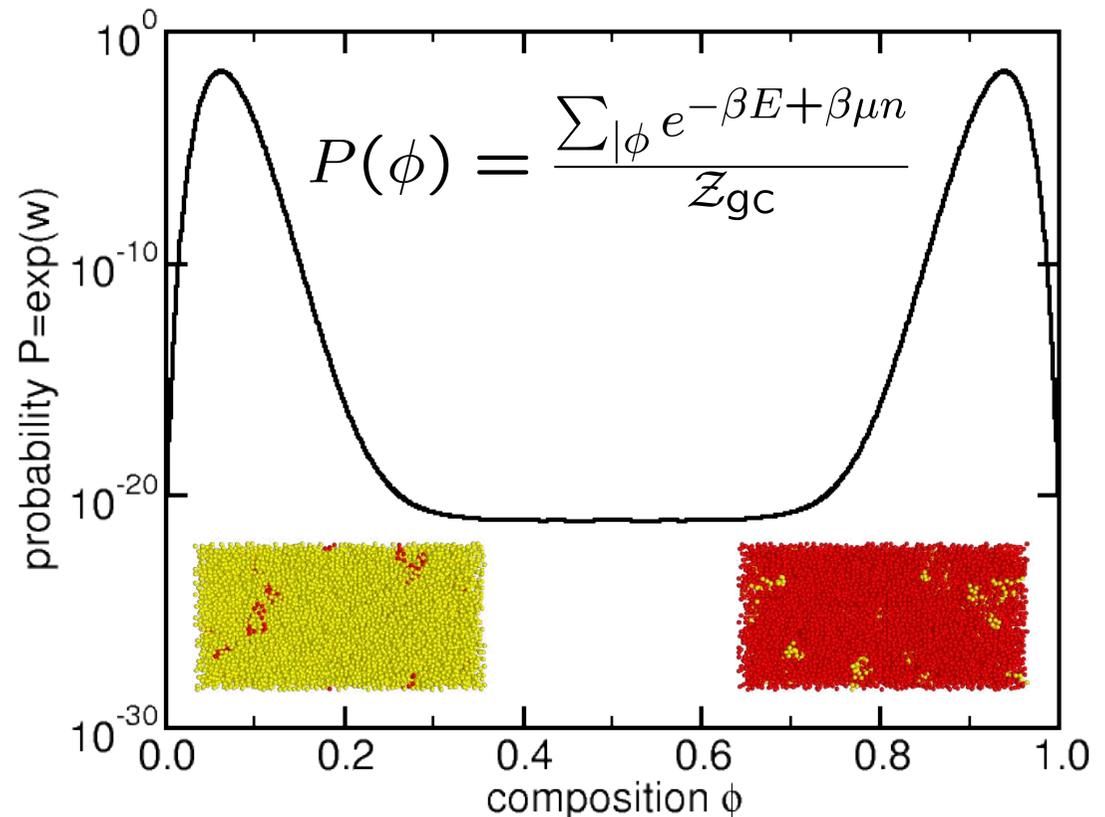
$$\mathcal{Z}_{\text{gc}} = \sum_{\{\text{conf}\}} e^{-\beta E + \beta \mu n}$$

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

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

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

➔ exponentially small finite-size effects



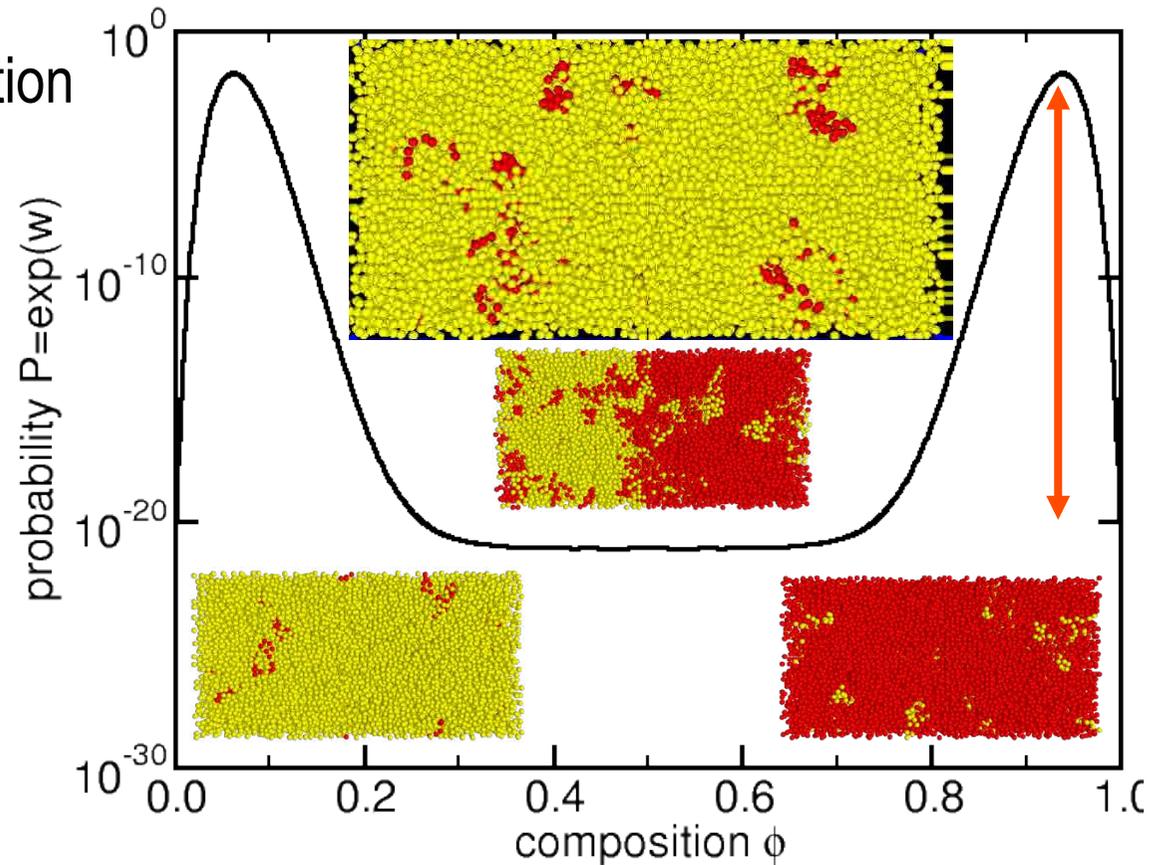
semi-grandcanonical simulations

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)

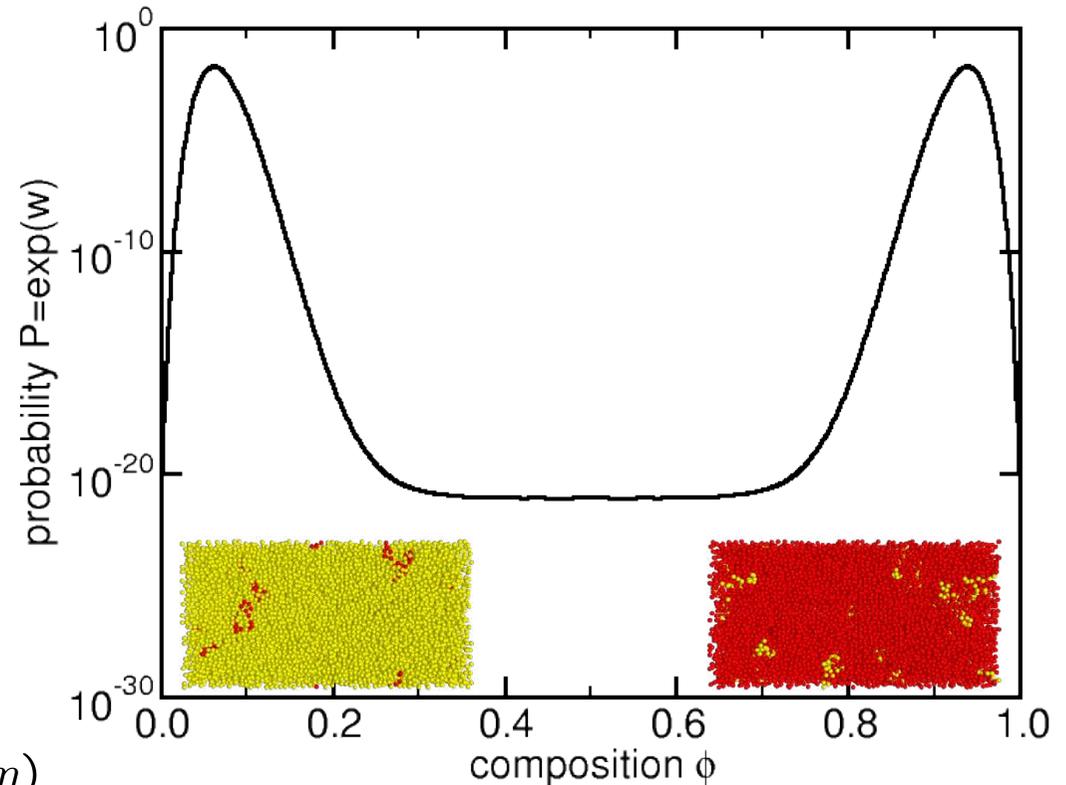
$$\mathcal{H}(\{r\}) = \mathcal{H}_0(\{r\}) + k_B T \eta(\phi)$$

additional term only depends on
composition ϕ 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 η every step

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

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

accumulate histogram

flat histogram?

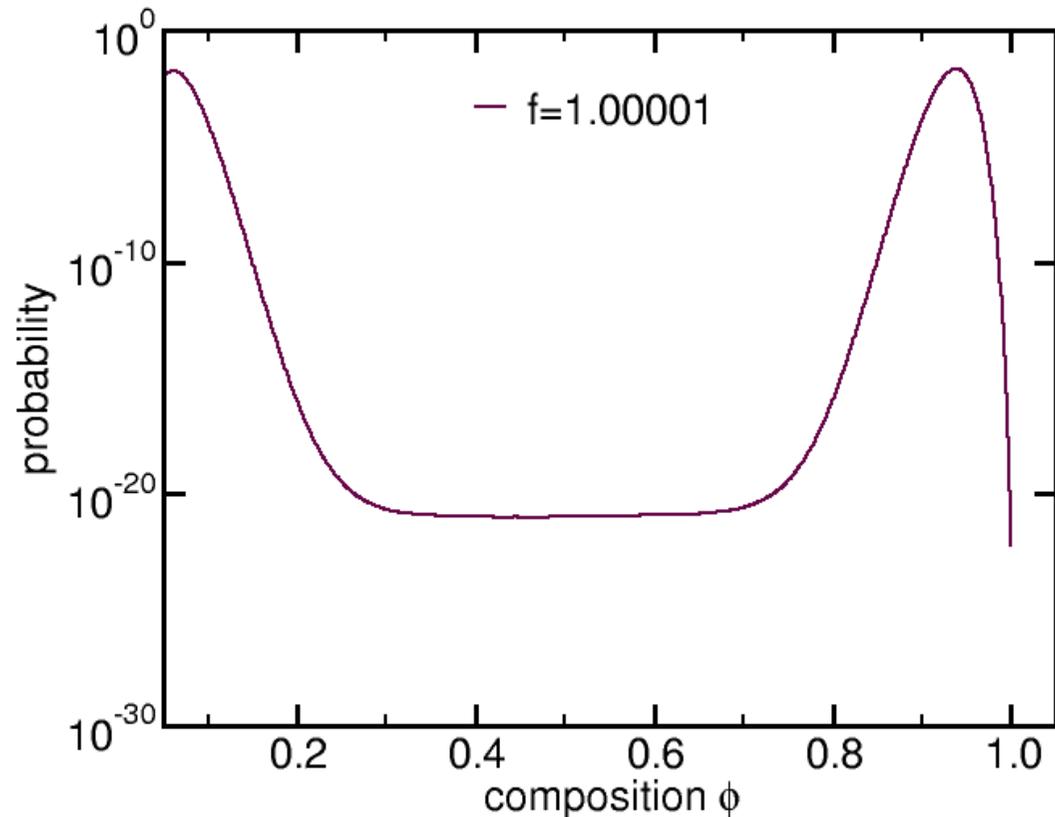
$$f \rightarrow \sqrt{f} \quad (\rightarrow 1)$$

reset histogram

but: no detailed balance for $f > 1$

errors difficult to estimate

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

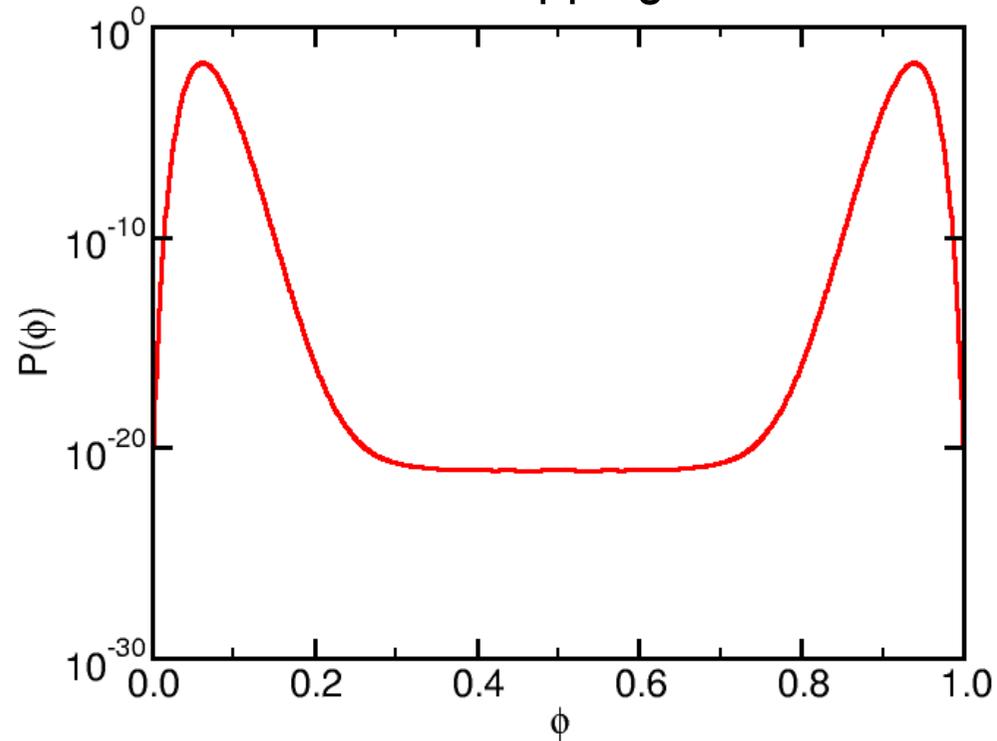


umbrella sampling

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

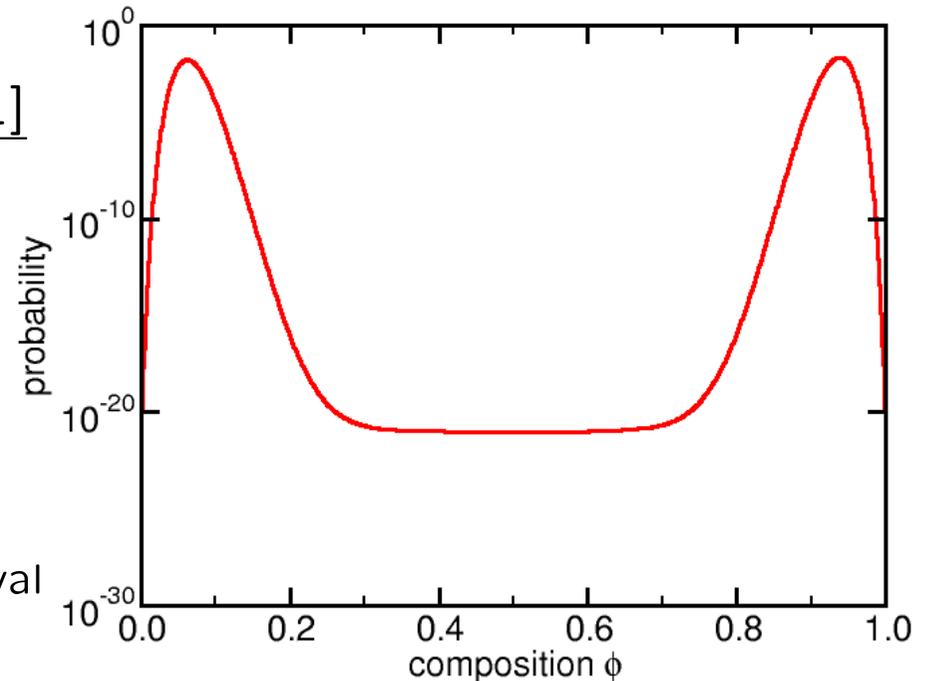
successive umbrella sampling

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

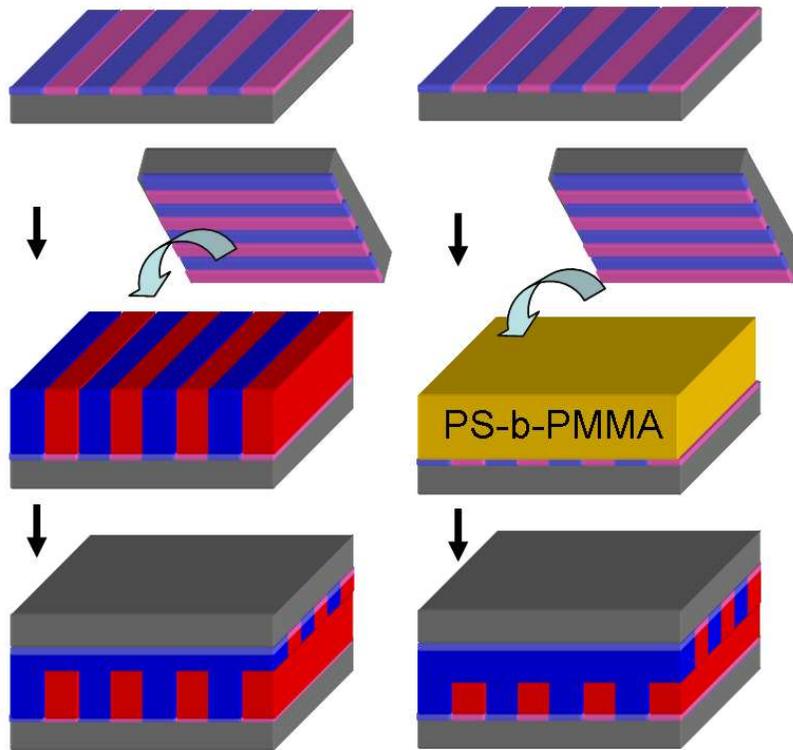
$$(\Delta P[n])^2 \sim \frac{1}{\text{tunnellings } 1 \leftrightarrow n} \sim \frac{n^2}{\#MCS}$$

- error allows for identification of sampling errors

➡ total error does not depend on the window size

interface localization delocalization-transition

morphology between apposing stripe patterns



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

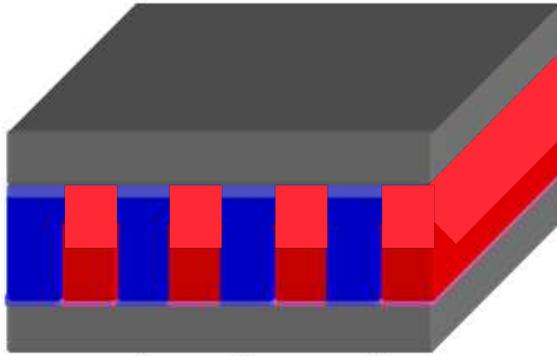
Liu, Ramirez-Hernandez, Yoshida, Nygard, Satapathy, Bunk, de Pablo, Nealey, PRL 108, 065502 (2012)

question: location of the grain boundary ?

➔ control of three-dimensional morphology

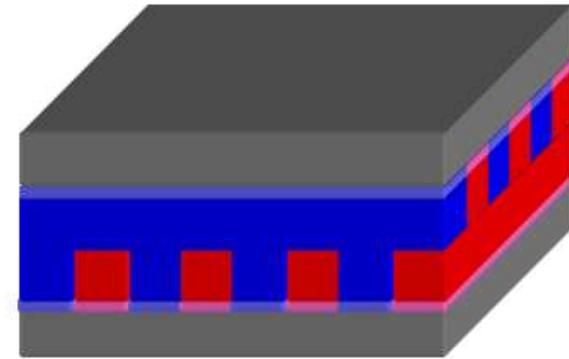
analogy between behavior of twist grain boundary and wetting in a film with strictly anti-symmetric boundaries (interface localization-delocalization)
control via geometric parameters of the system

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:

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

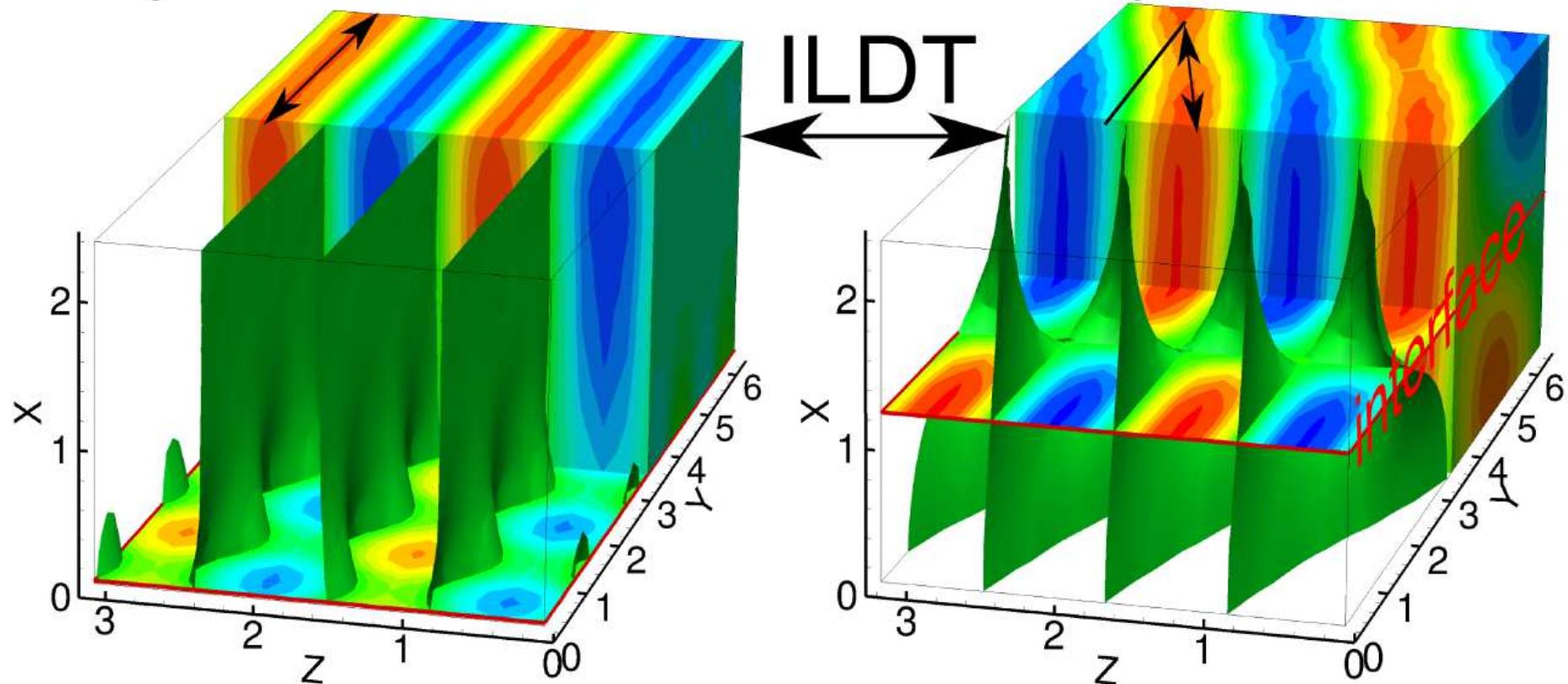
$$+ \frac{1}{2}B(\lambda_b - \lambda_0)^2x + \frac{1}{2}B(\lambda_t - \lambda_0)^2(D - x) \quad \text{stretch/compression } (\mu)$$

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

interface localization delocalization-transition

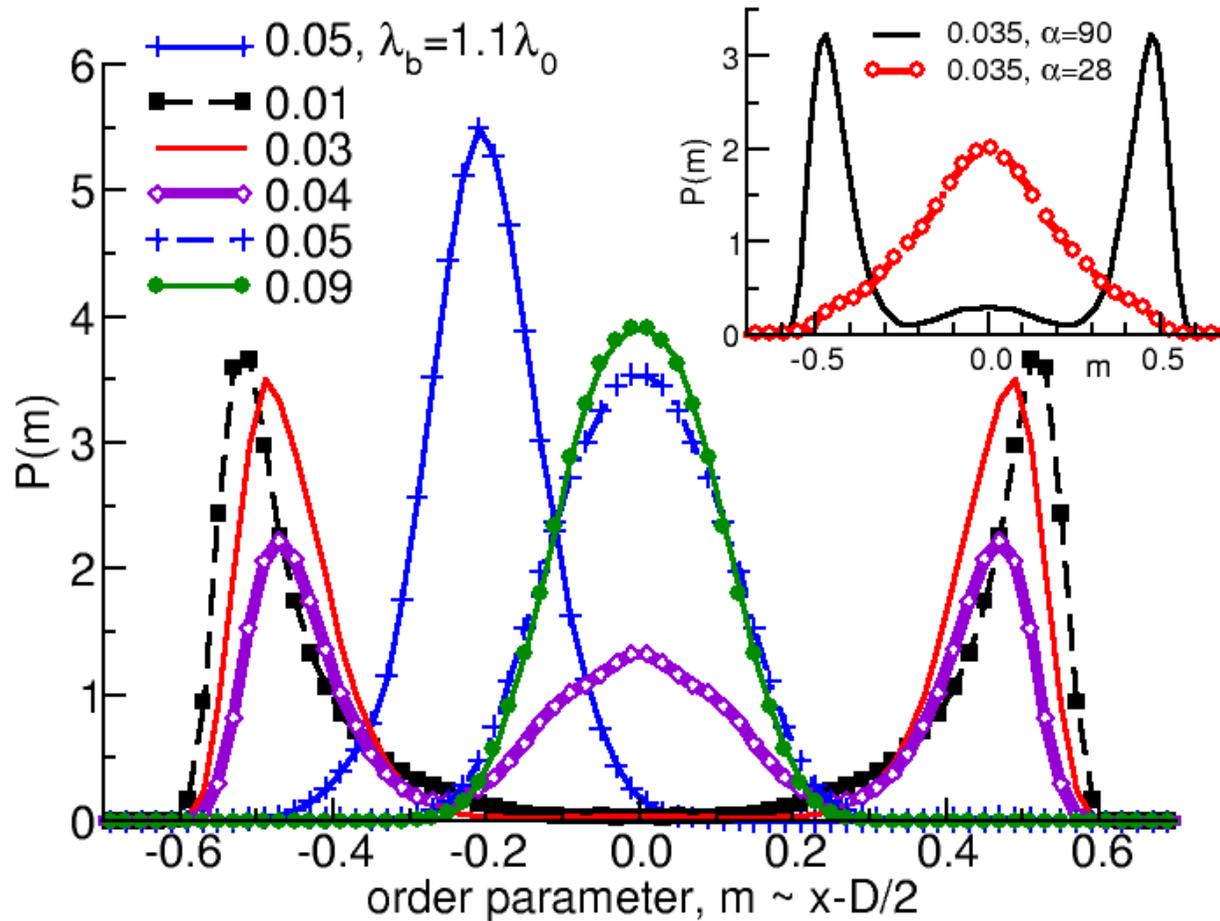
twist angle $\alpha=90^\circ$, localized at bottom

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



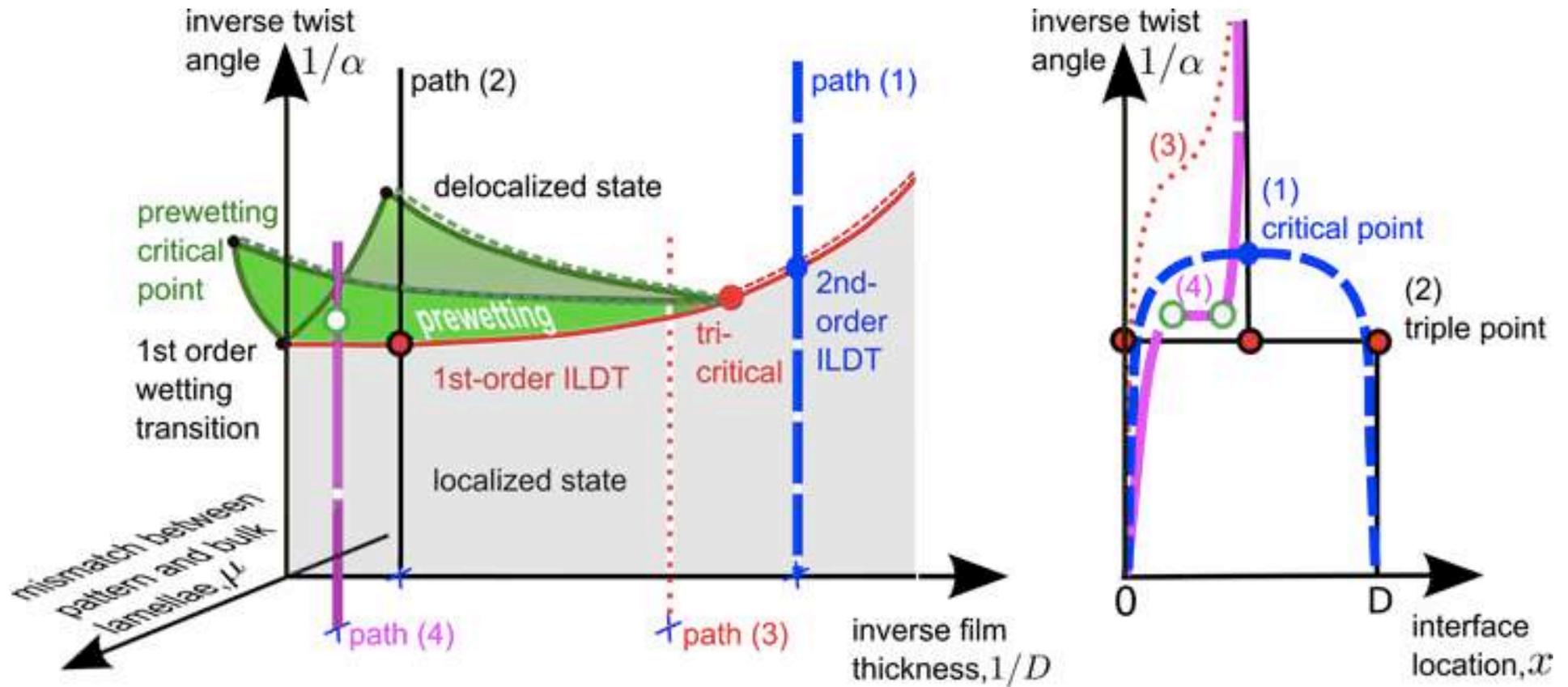
$$\frac{\Delta F(x)}{A} = \gamma_b(\Lambda) + \gamma_t(\Lambda) + \gamma(\alpha)$$

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

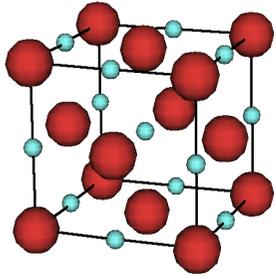
$$+ \frac{1}{2}B(\lambda_b - \lambda_0)^2 x + \frac{1}{2}B(\lambda_t - \lambda_0)^2 (D - x)$$

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

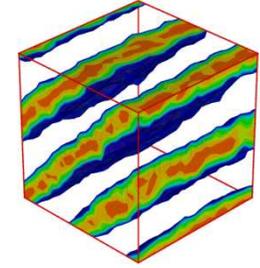
interface localization delocalization-transition



- 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

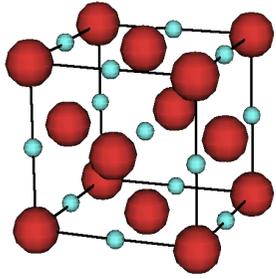
Einstein crystal is reference state
use thermodynamic integration wrt
to uniform, harmonic coupling of
particles to ideal position

(Frenkel & Ladd)

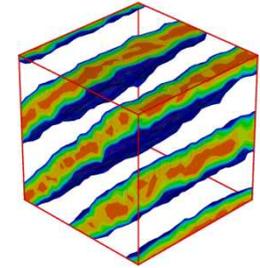
Fourier mode of composition fluctuation
SCFT solution
homogeneous fluid/melt

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

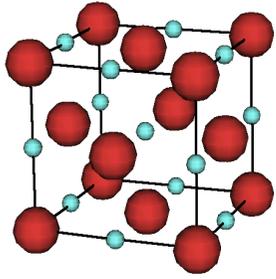
ordered state: particles vibrate
 around ideal lattice positions

Fourier mode of composition fluctuation
 SCFT solution
 homogeneous fluid/melt

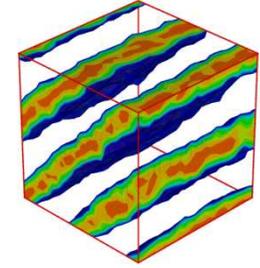
ordered phase: composition fluctuates
 around reference state (SCFT solution),
 but **molecules diffuse** (liquid)

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

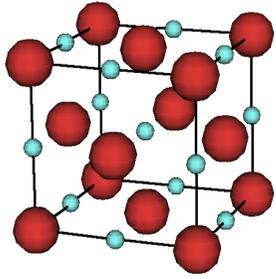
ordered state: particles vibrate
around ideal lattice positions

Fourier mode of composition fluctuation
SCFT solution
homogeneous fluid/melt

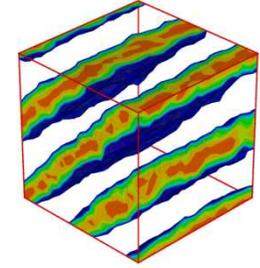
ordered phase: composition fluctuates
around reference state (SCFT solution),
but ***molecules diffuse*** (liquid)

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}
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

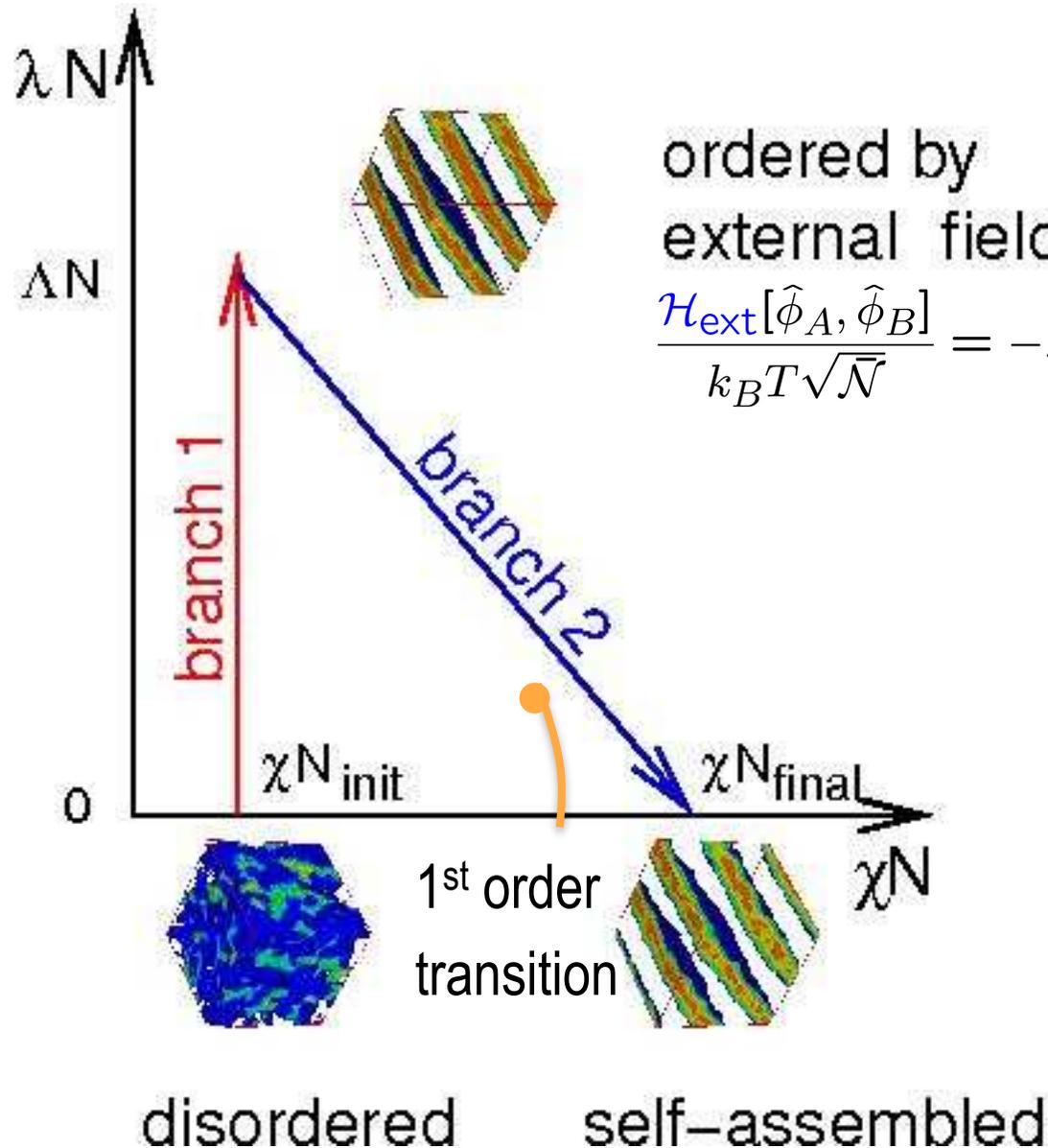
PRE 51, R3795 (1995)

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

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

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)

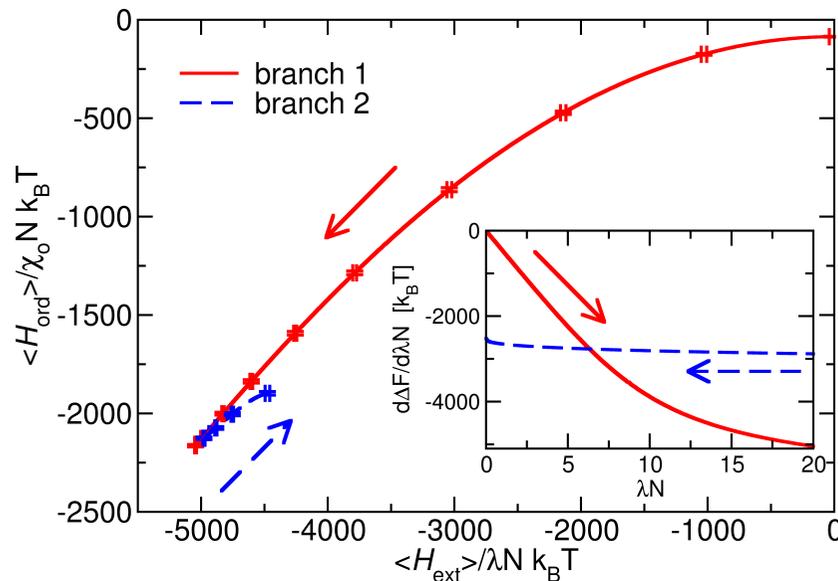
free energy difference via TDI

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

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

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

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



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

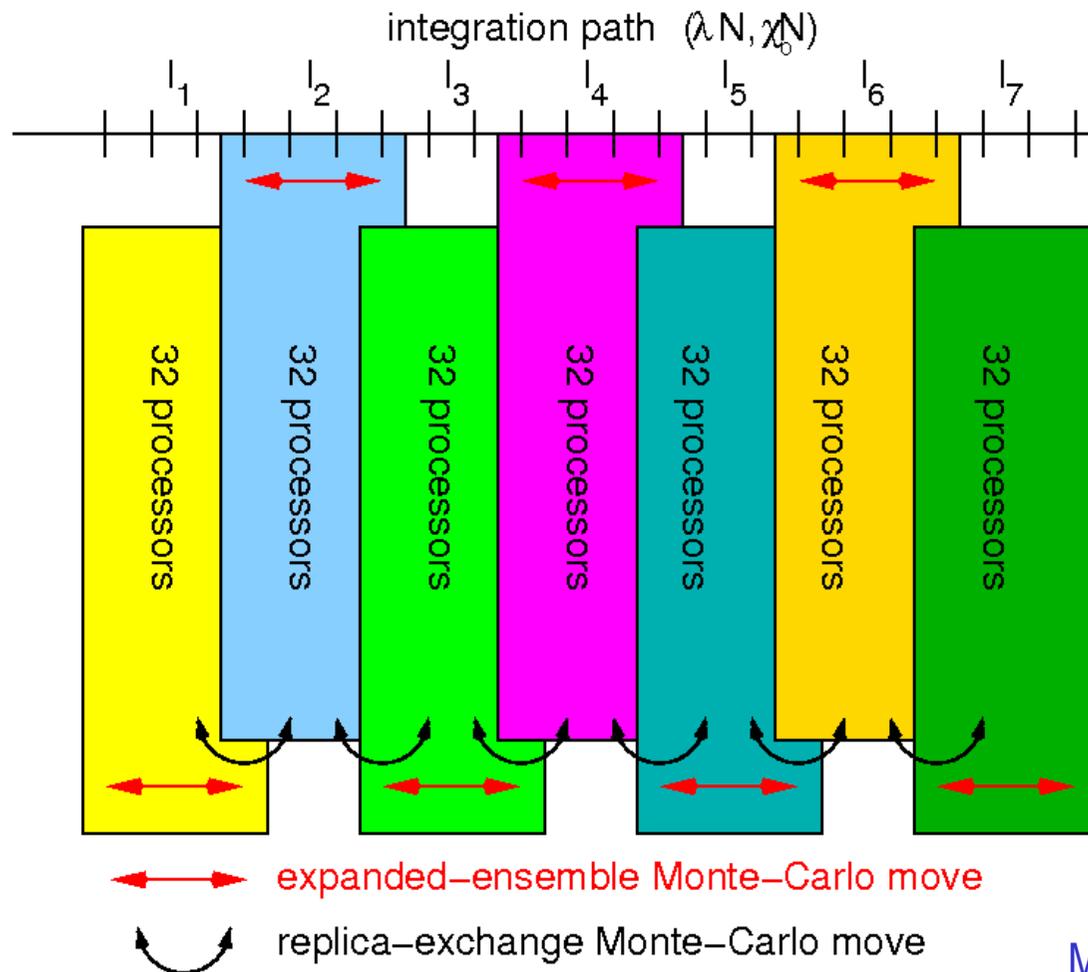
$$-\lambda N h_{\text{eff}}(\mathbf{r}) = \frac{V}{nk_B T} \left(\frac{\delta \mathcal{H}_{\text{ord}}}{\delta \phi(\mathbf{r})} + \frac{\delta \mathcal{H}_{\text{ext}}}{\delta \phi(\mathbf{r})} \right)$$

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

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

TDI vs expanded ensemble/replica exchange

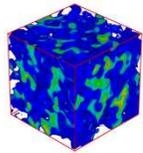
$$\mathcal{Z} \sim \prod_{j=1}^{M_r} \sum_{\lambda_i N \in I_j} \frac{e^{\eta(\lambda_i N)}}{n!} \int \prod_{i=1}^n \mathcal{D}[\mathbf{r}_i(s)] e^{-\frac{\mathcal{H}_b + \mathcal{H}_{\text{liq}} + \mathcal{H}_{\text{ord}} + \mathcal{H}_{\text{ext}}}{k_B T}}$$



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

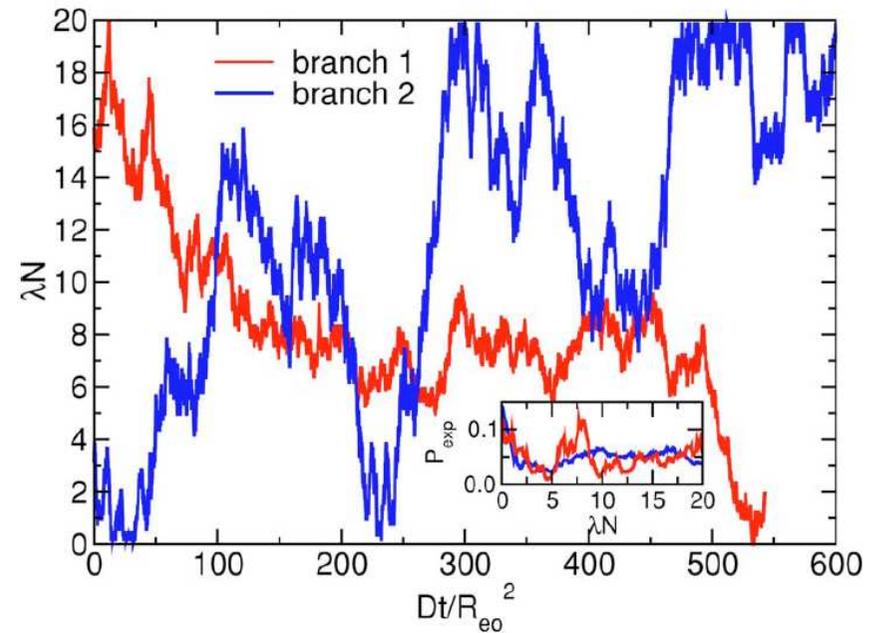
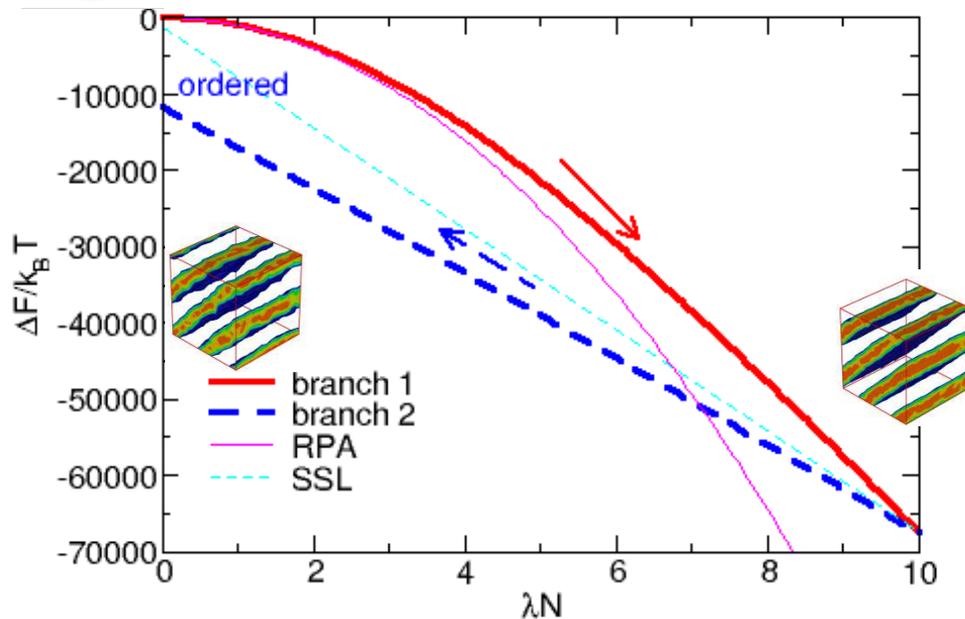
accuracy of the method

no kinetic barrier, ie no phase transition
roughly equal probability



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

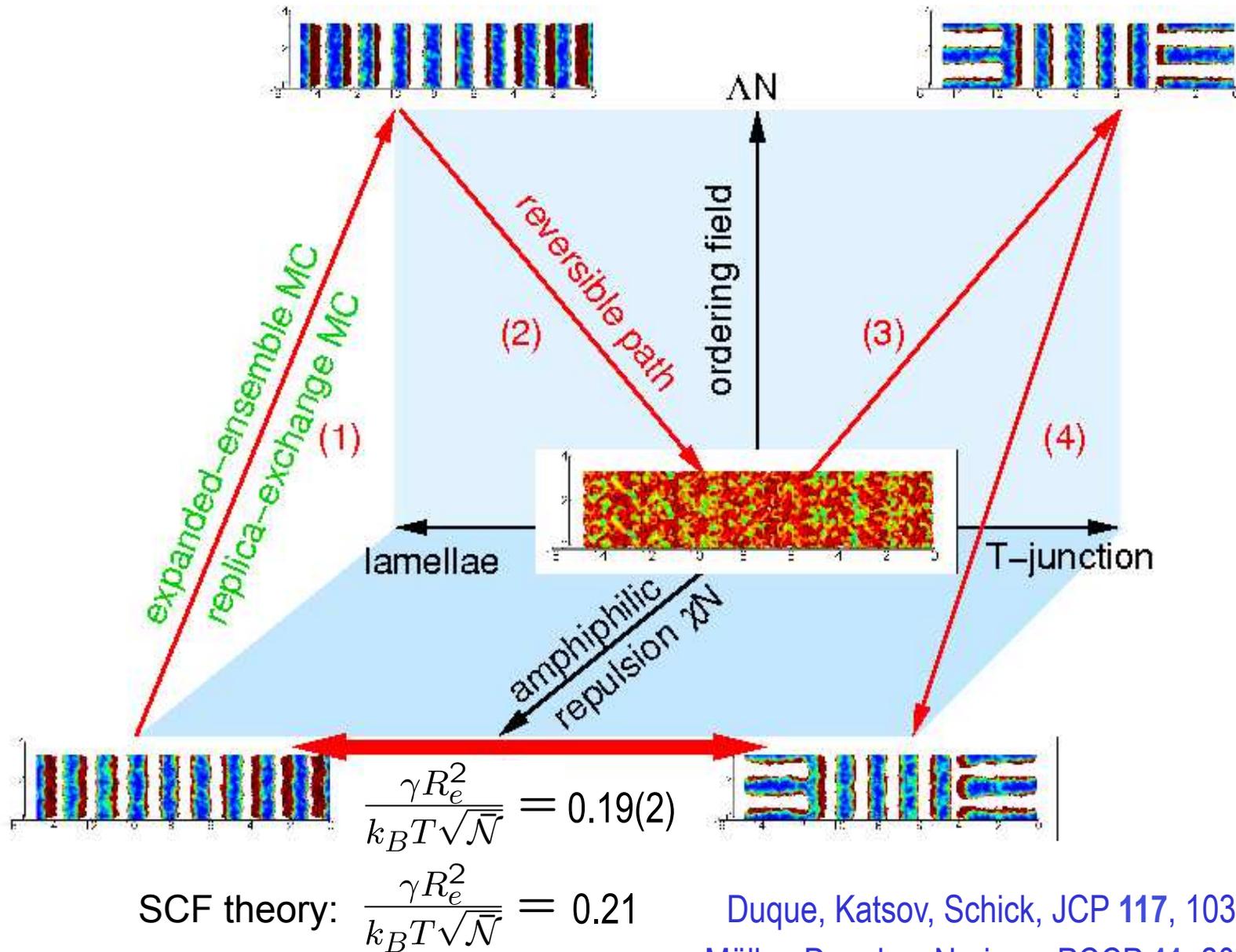
$\delta \Delta F = \mathcal{O}(10^1)k_B T$



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

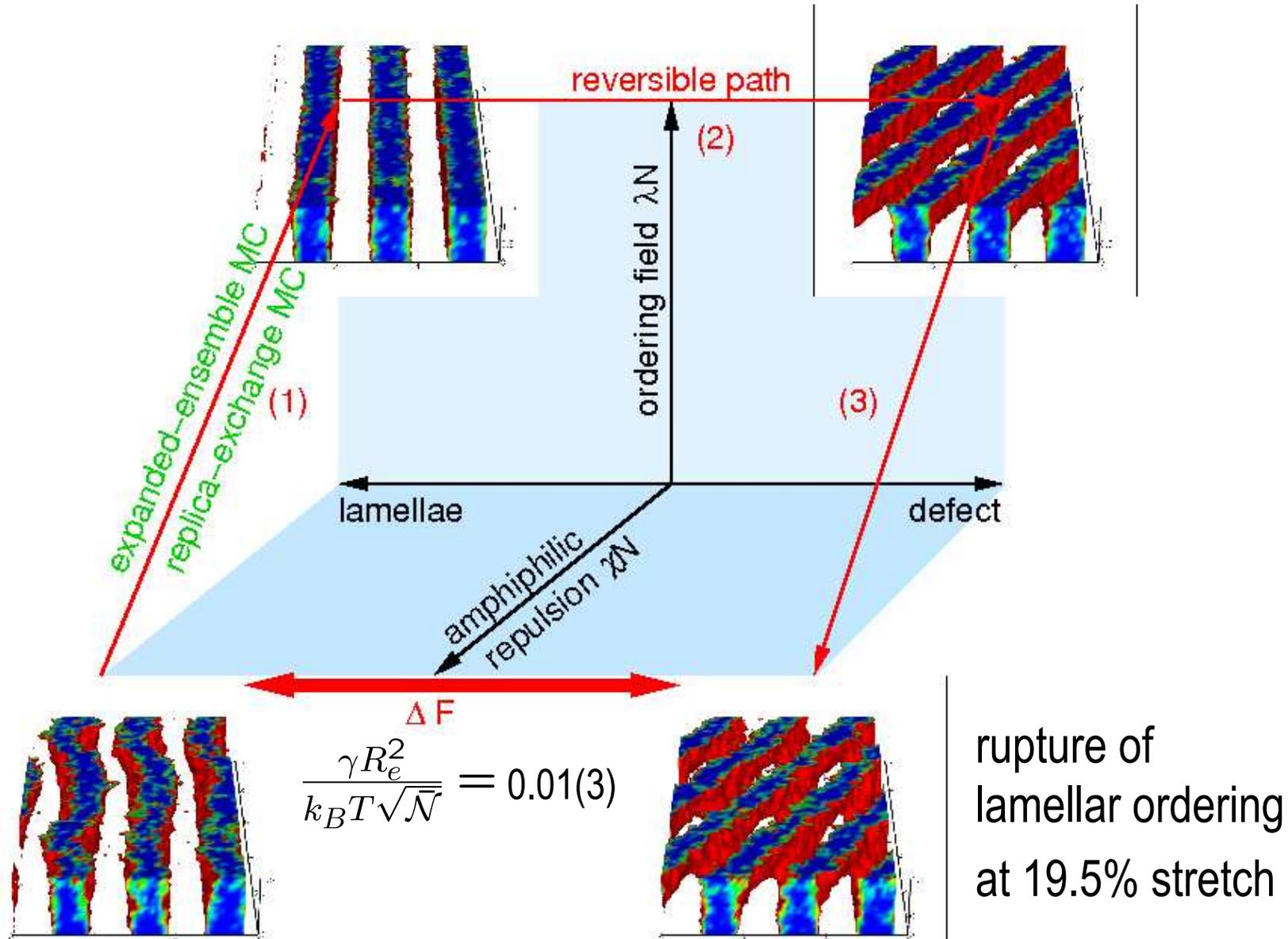
reweighting technique removes large free energy change along the path
 probability distribution of reweighted simulation estimates accuracy
 kinetics demonstrates the absence of first-order transition

free-energy calculations: grain boundaries

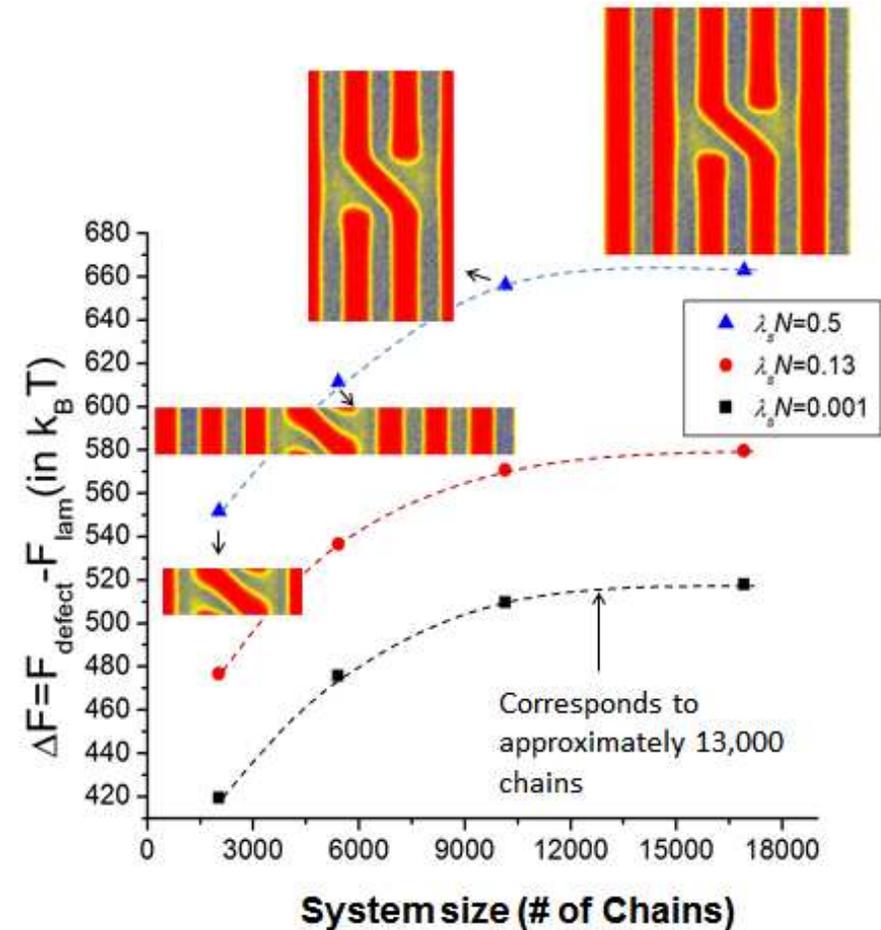
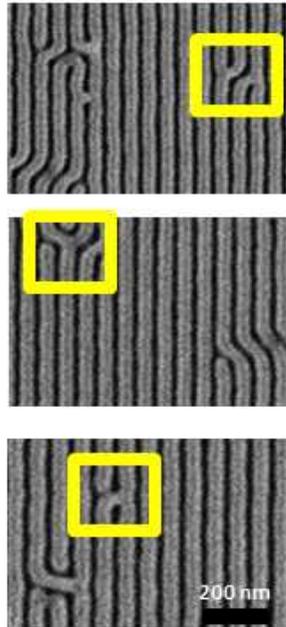
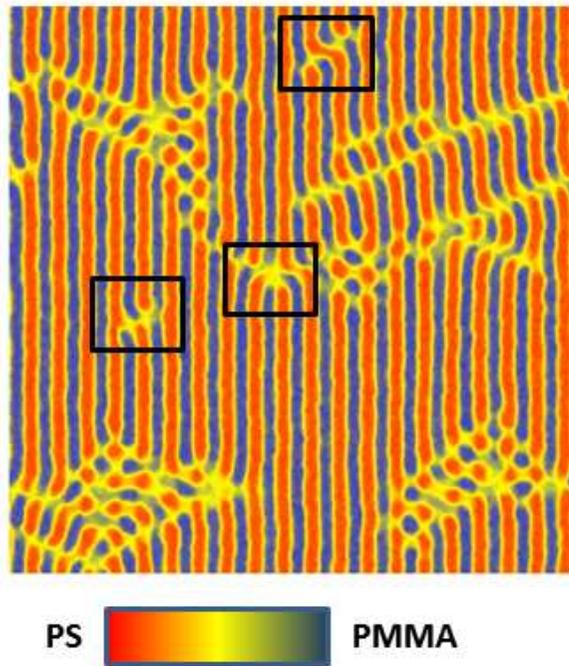


Duque, Katsov, Schick, JCP 117, 10315 (2002)
 Müller, Daoulas, Norizoe, PCCP 11, 2087 (2009)

reconstruction of soft morphology at patterned surface



defect free energy for lamellar pattern replication



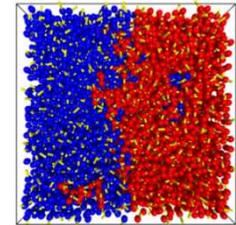
naïve estimate:

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

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), ∞

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

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

model definition:

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

free-energy functional, $\mathcal{F}_{GL}[m(\mathbf{r})]$
 (Ginzburg-Landau-de Gennes or Ohta-Kawasaki)

(here: Rouse dynamics)

time-dependent GL theory

segmental friction, ζ

(model B according to Hohenberg & Halperin)

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

projection:

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

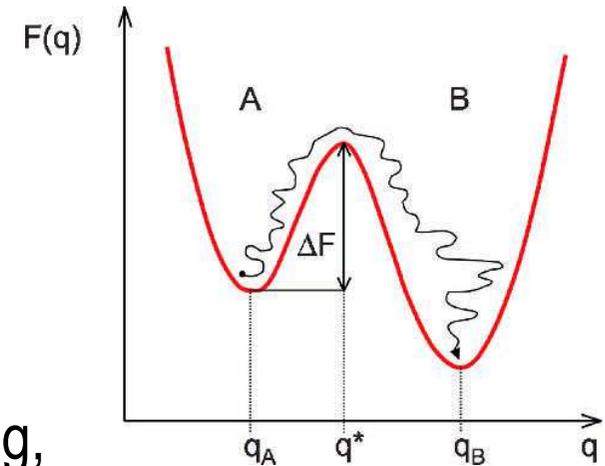
$$\frac{\mathcal{F}[m]}{k_B T} \equiv - \ln \int \mathcal{D}[\{\mathbf{r}_{i,s}\}] e^{-\frac{\mathcal{H}(\{\mathbf{r}_{i,s}\})}{k_B T}} \delta[m - (\hat{\phi}_A - \hat{\phi}_B)]$$

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

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)

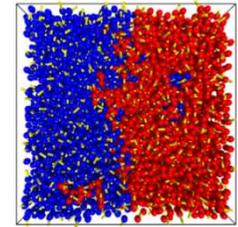
•SCMF simulation

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

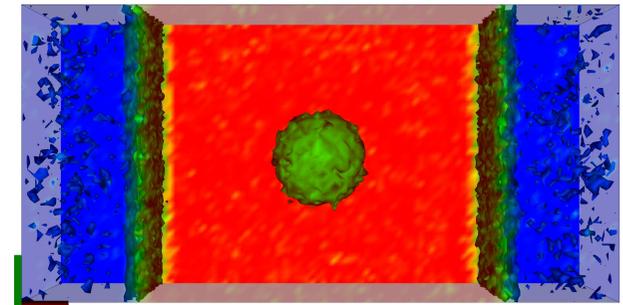
- **Lifshitz-Slyozov coarsening** (diffusive regime)

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

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

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

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



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

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

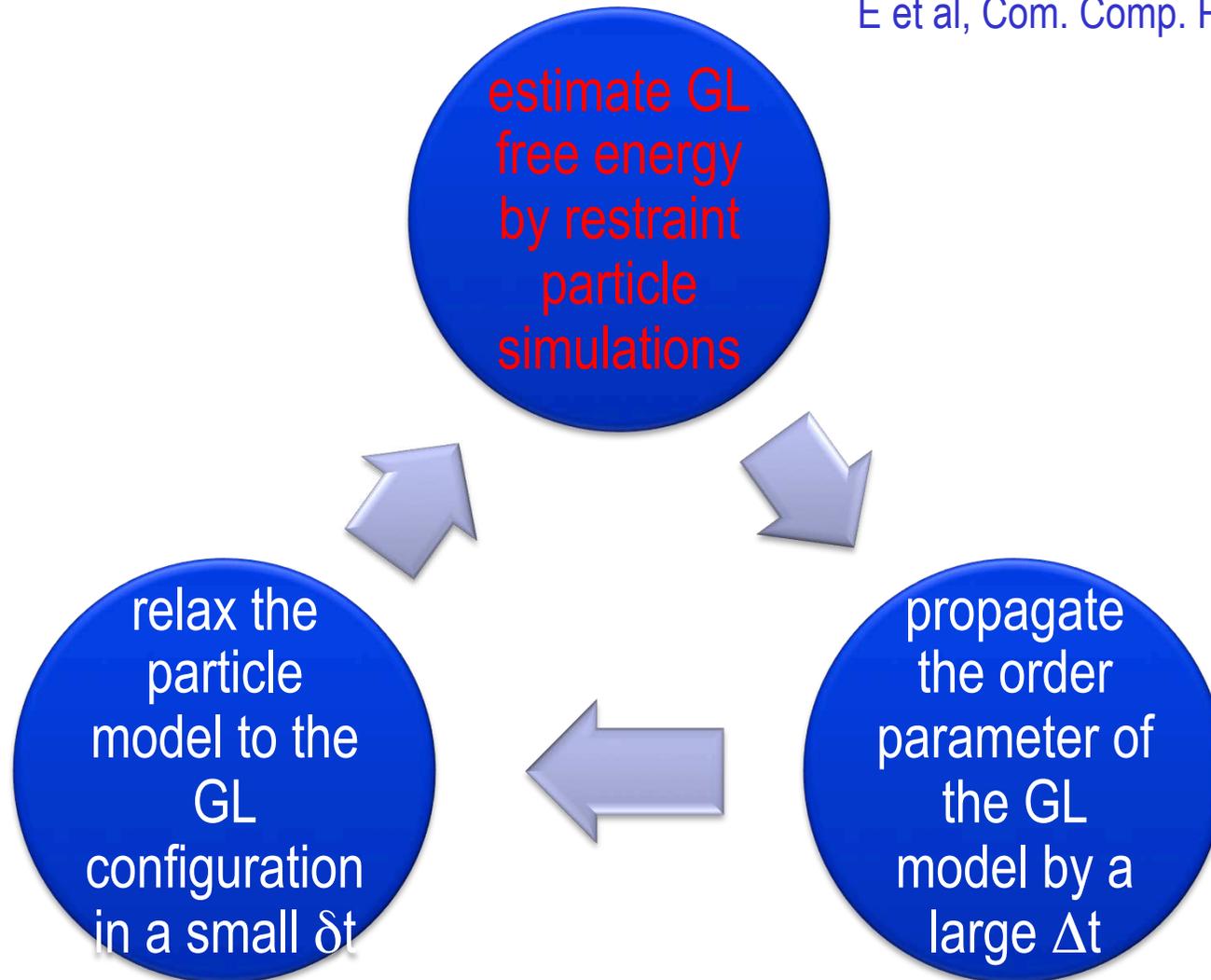
- alternative: droplet coagulation by Brownian motion

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

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, $m(\mathbf{r})$)

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

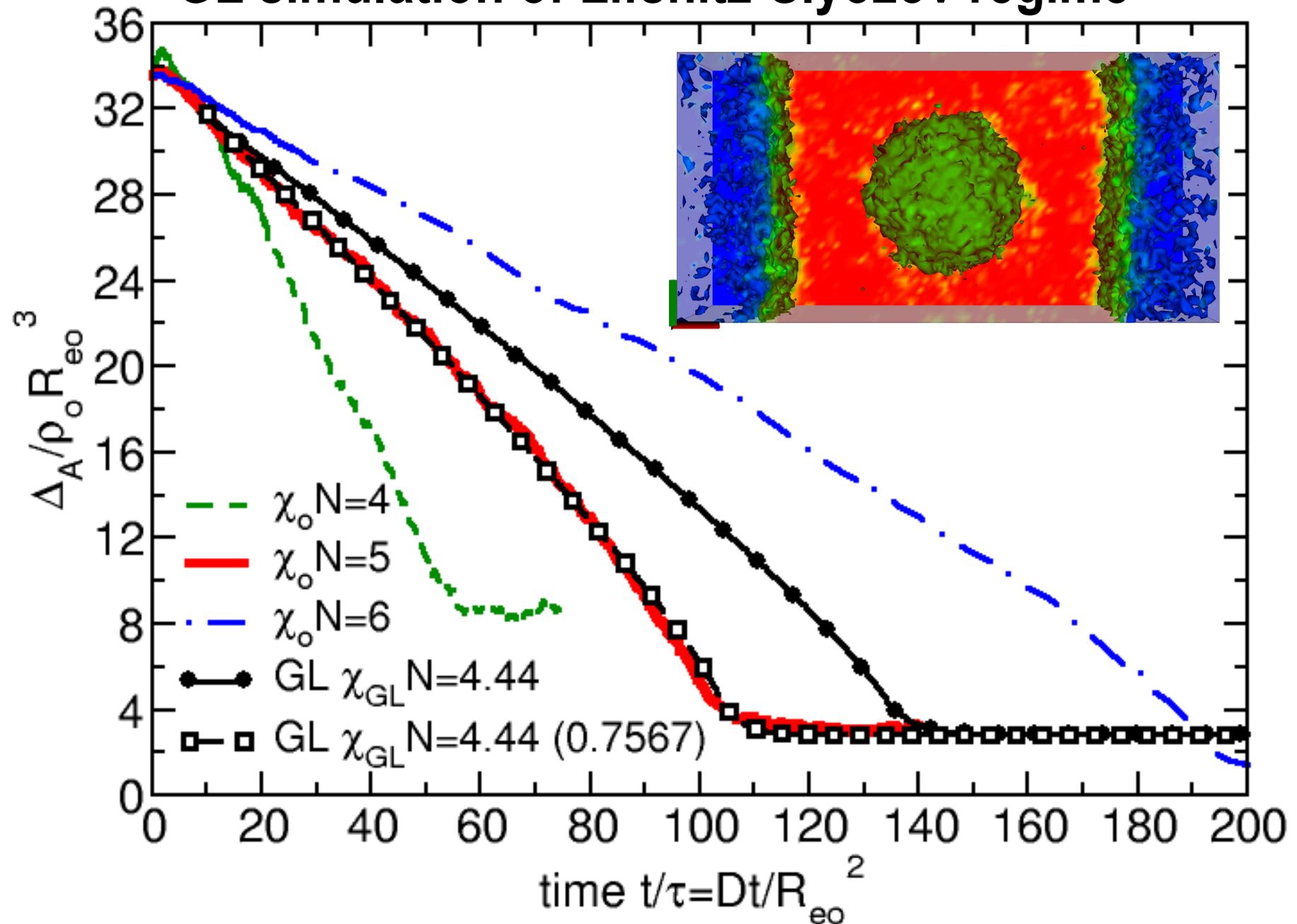
GL *model* required, not “equation-free” Kevrekidis, Gear, Hummer *AICHE J.* **50**, 1346 (2004)
average over space (instead of time) to determine the few parameters
of the Ginzburg-Landau model (spatial homogeneity of GL model)

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

large time step Δt is limited by the condition that parameters do not vary on scale Δt

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

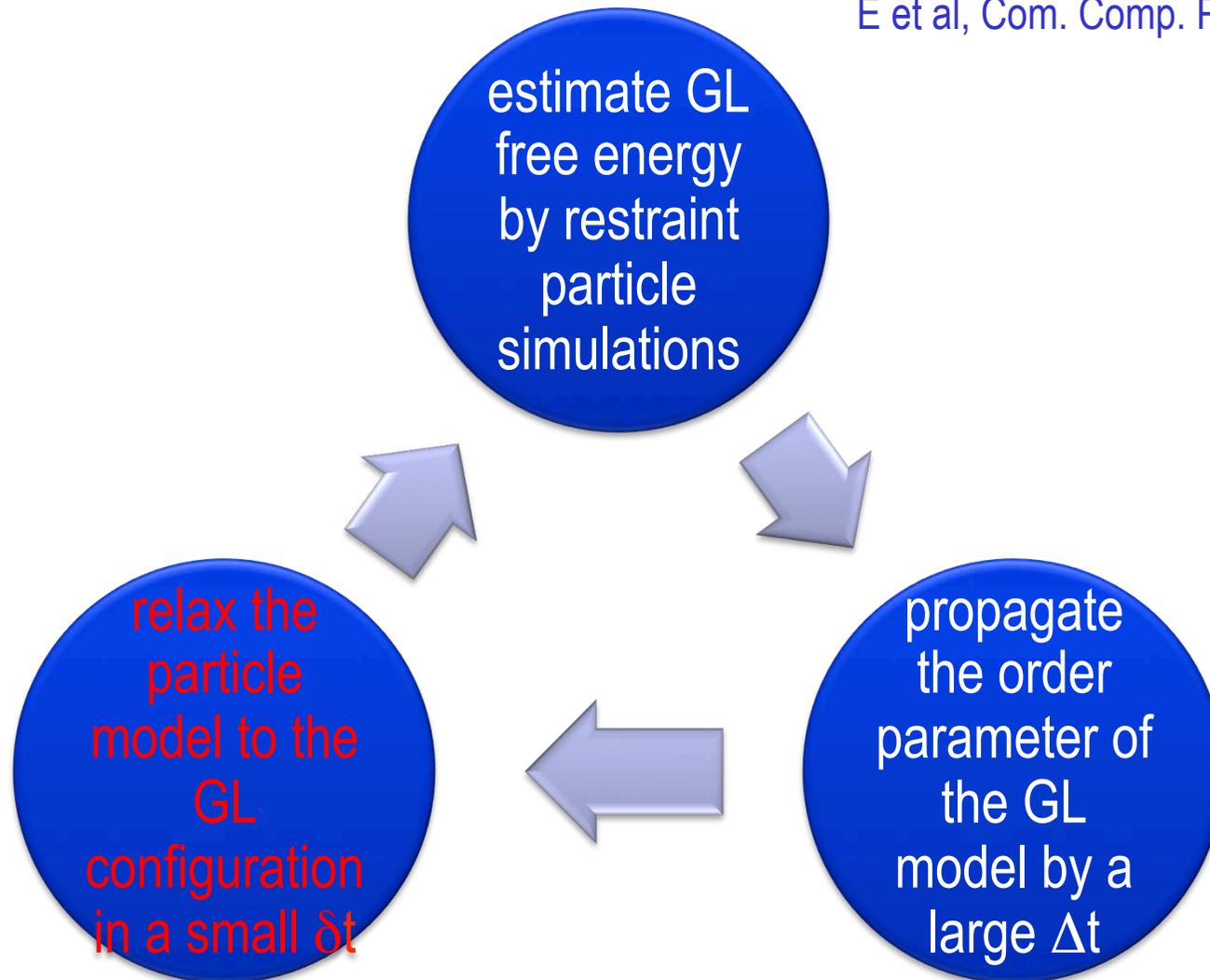
GL simulation of Lifshitz-Slyozov regime



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 $\frac{\tau}{\lambda N}$

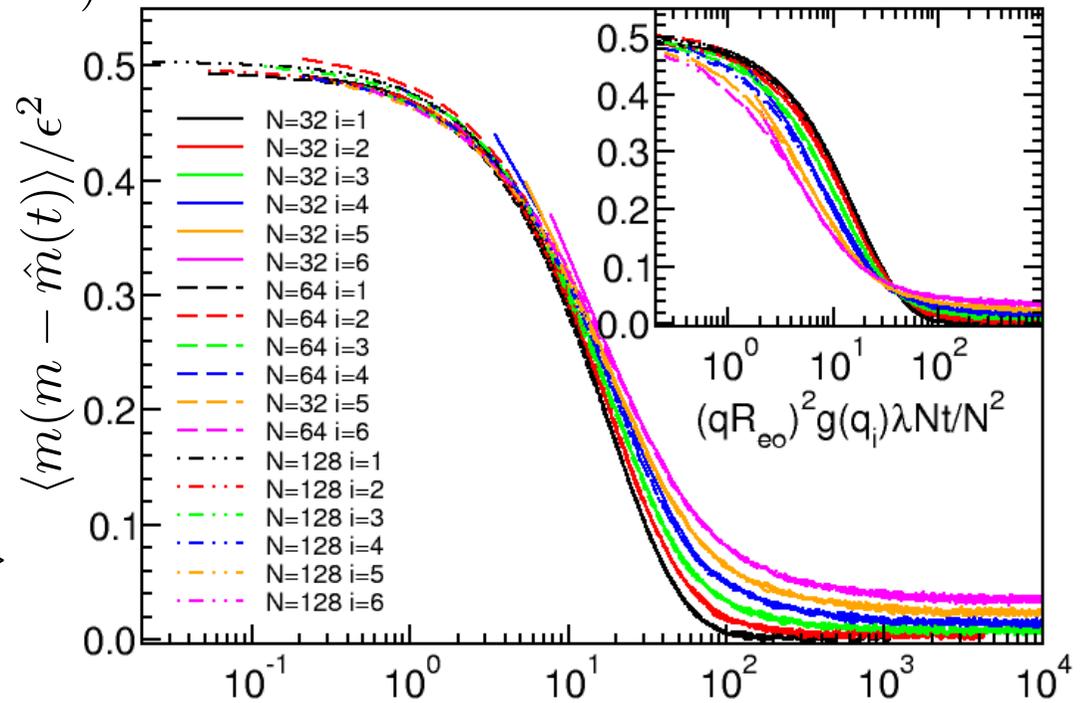
Onsager coefficient from relaxation of simulations

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

relaxation time of the constraint system is speeded-up by a factor λN

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

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



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

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

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

➔ 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 λN ?

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

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

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

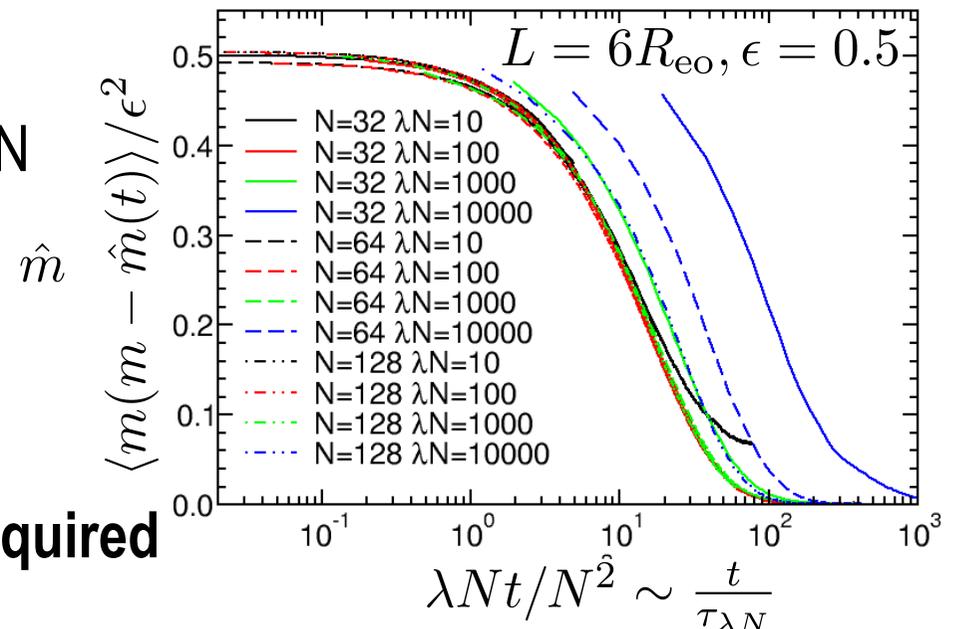
$$f_{\lambda N} \sim \frac{k_B T \lambda}{\Delta L} \sim \frac{k_B T \lambda N}{R_{eo}} N^{-1/2}$$

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

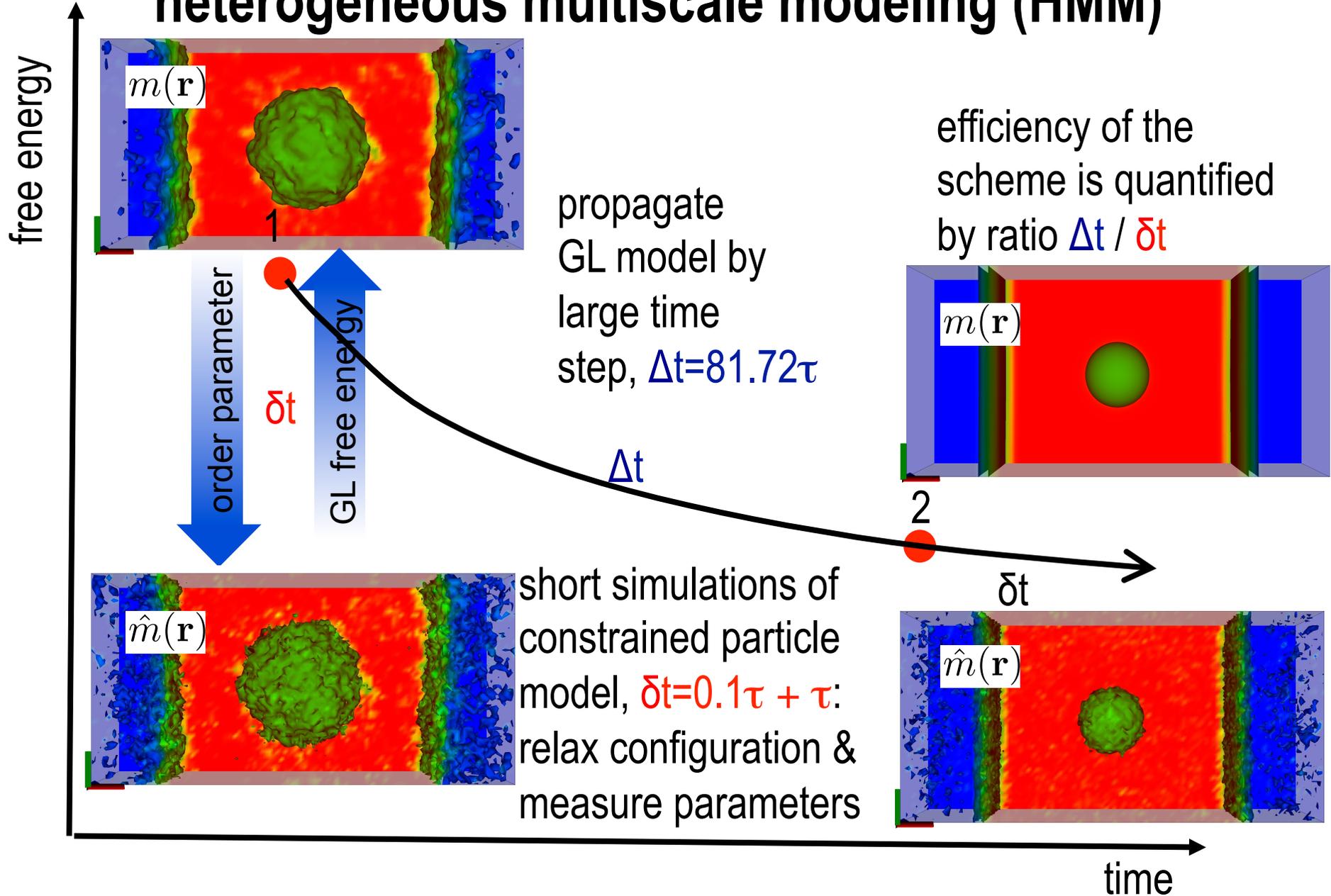
relaxation rate is increase by a factor λN
 for small composition/density variations

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

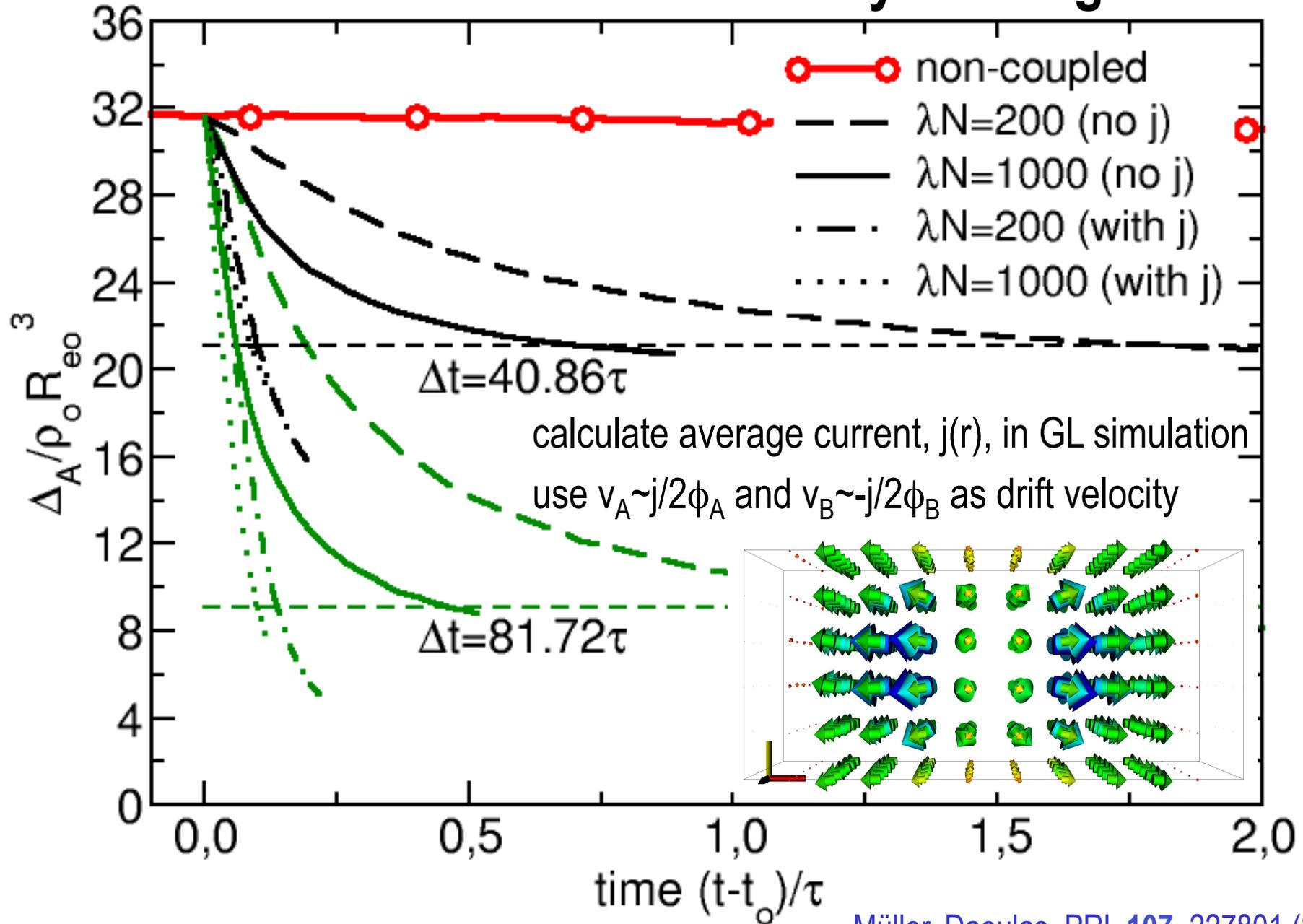
additional relaxation τ may be required



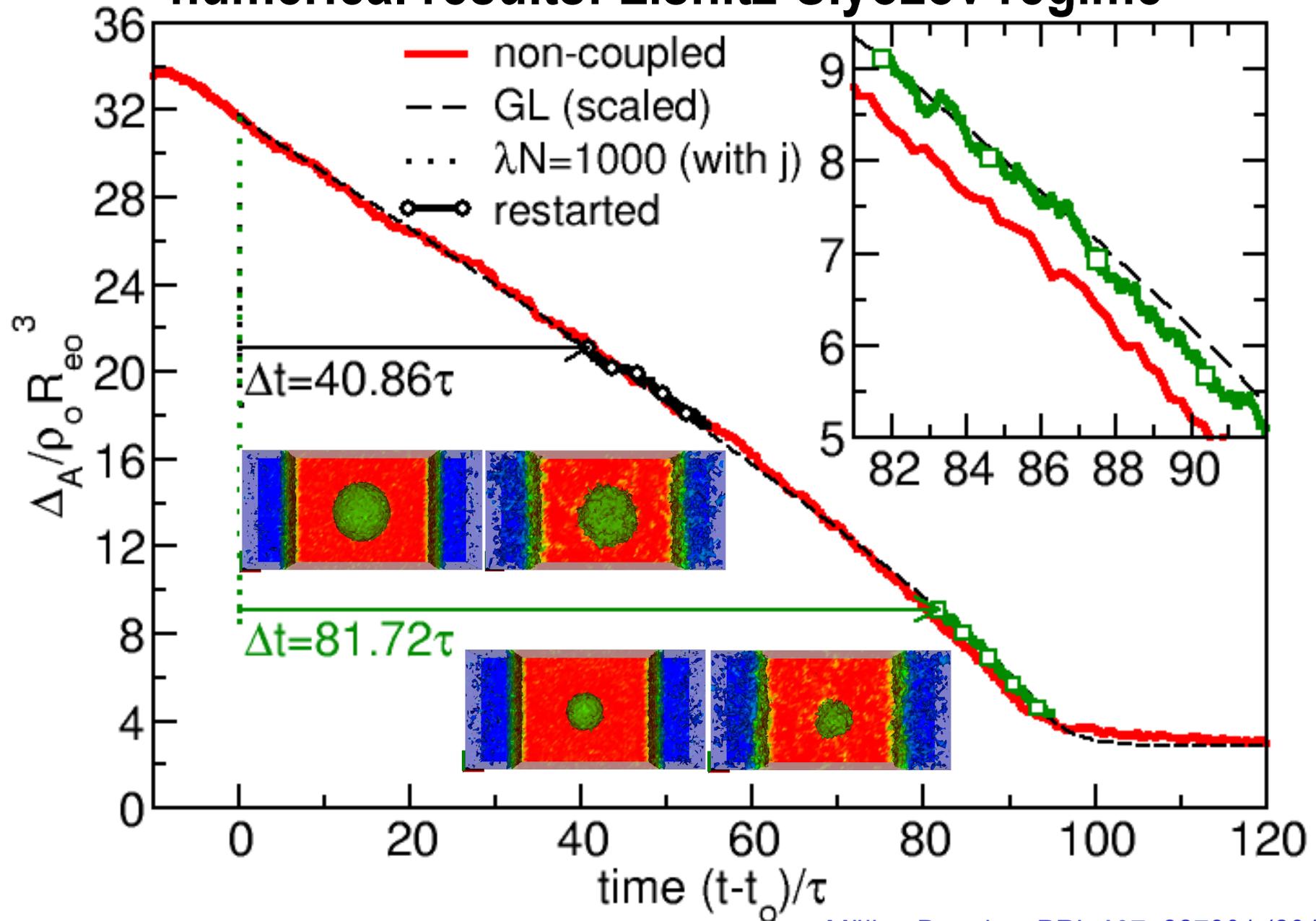
heterogeneous multiscale modeling (HMM)



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

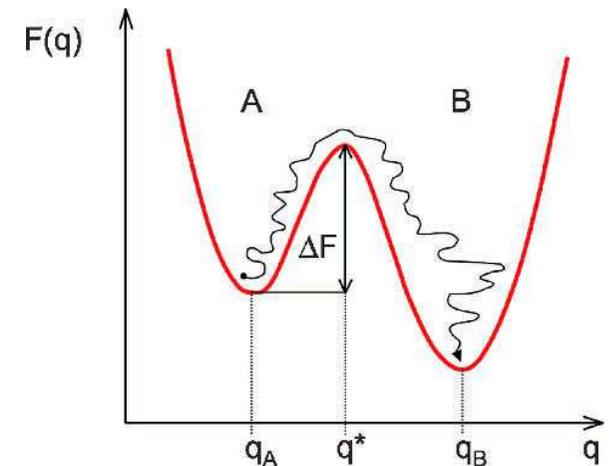
compute free-energy functional

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

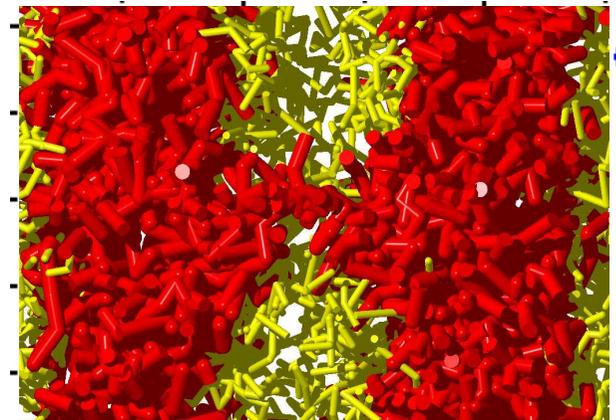
➔ field-theoretic umbrella sampling,
on-the-fly string method (particle-based)

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

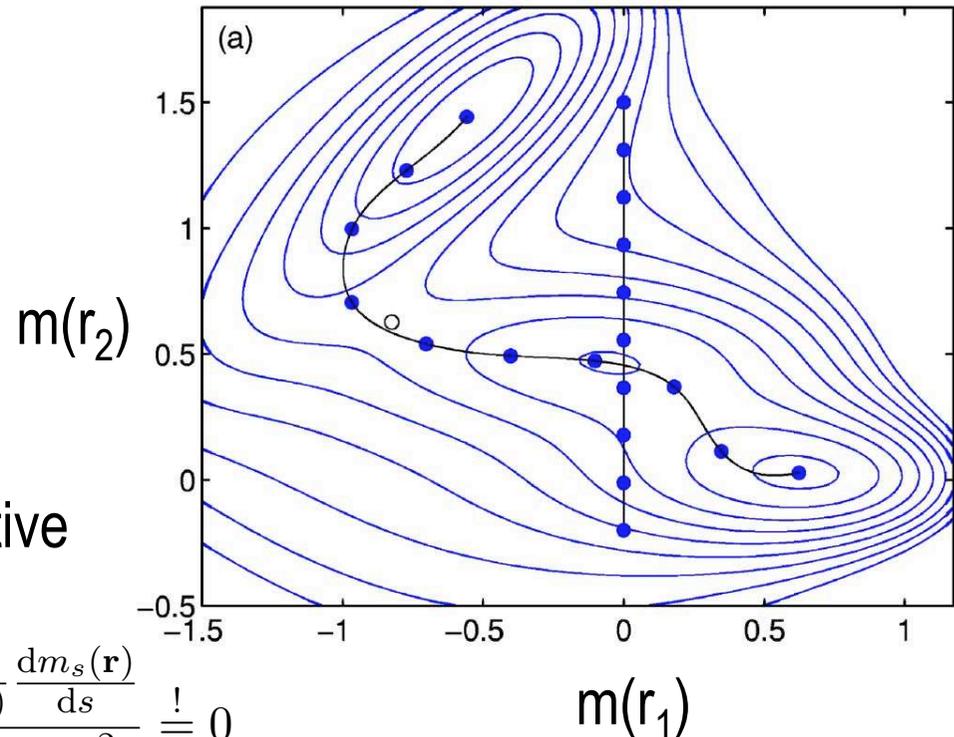
$$\nabla_{\perp} \mathcal{F}[m] = \frac{\delta \mathcal{F}}{\delta m_s(\mathbf{r})} - \frac{dm_s(\mathbf{r})}{ds} \frac{\int d^3\mathbf{r} \frac{\delta \mathcal{F}}{\delta m_s(\mathbf{r})} \frac{dm_s(\mathbf{r})}{ds}}{\int d^3\mathbf{r} \left(\frac{dm_s(\mathbf{r})}{ds}\right)^2} \stackrel{!}{=} 0$$

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

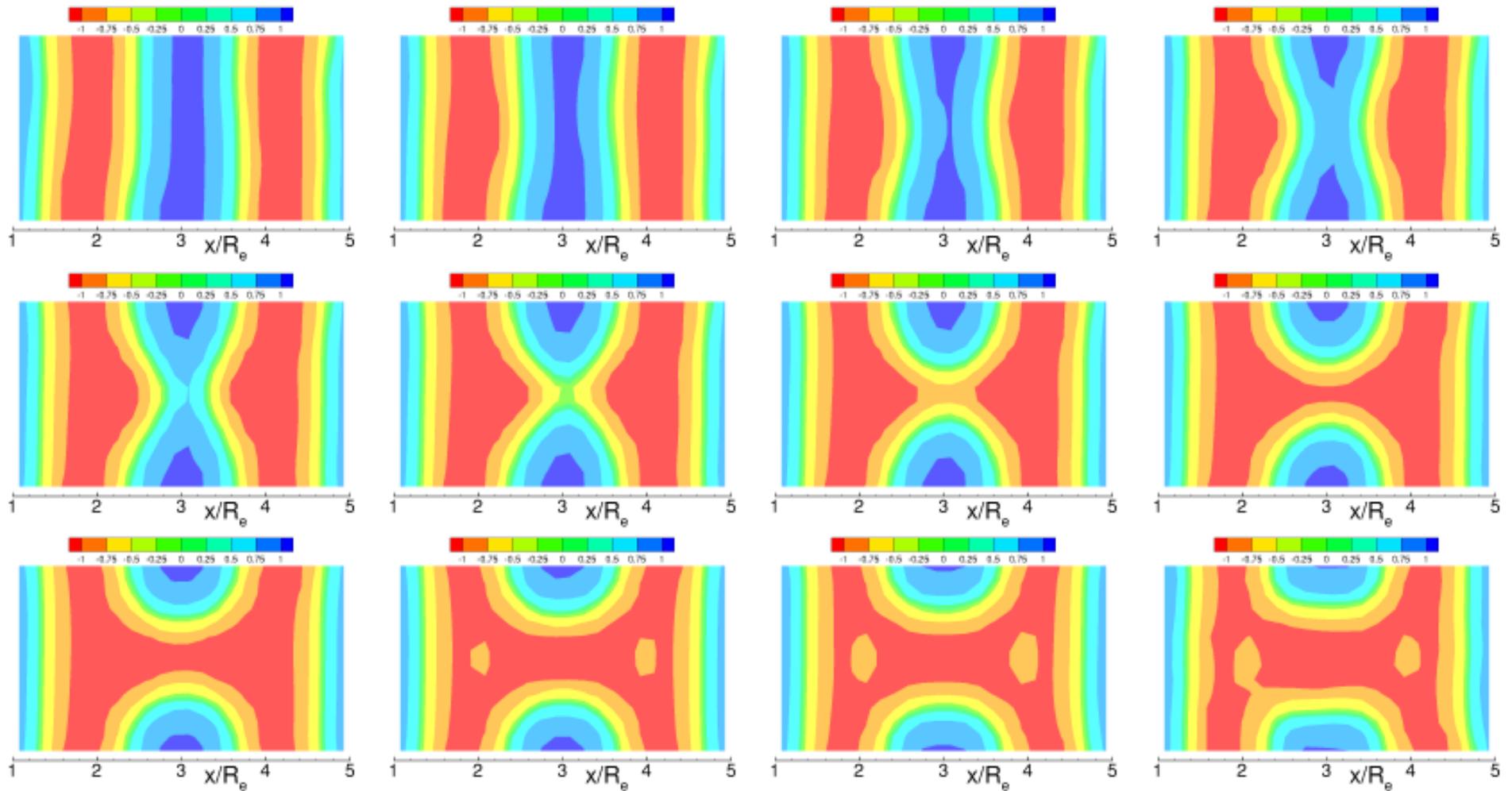
evolve each morphology $m_s(\mathbf{r})$ as to minimize the free energy

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

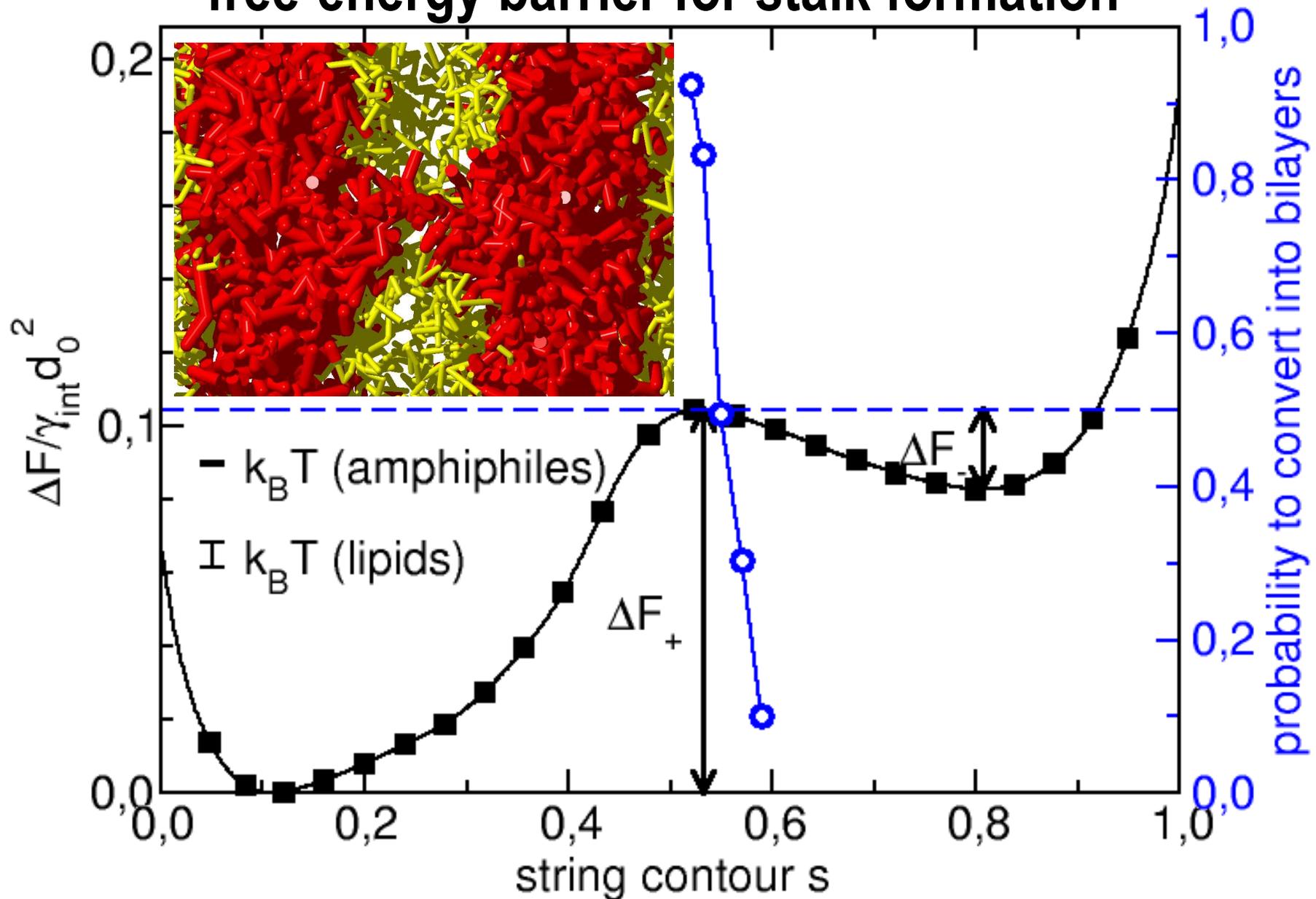
re-parameterize the string to equal distance Δs (pointwise 3rd order spline)



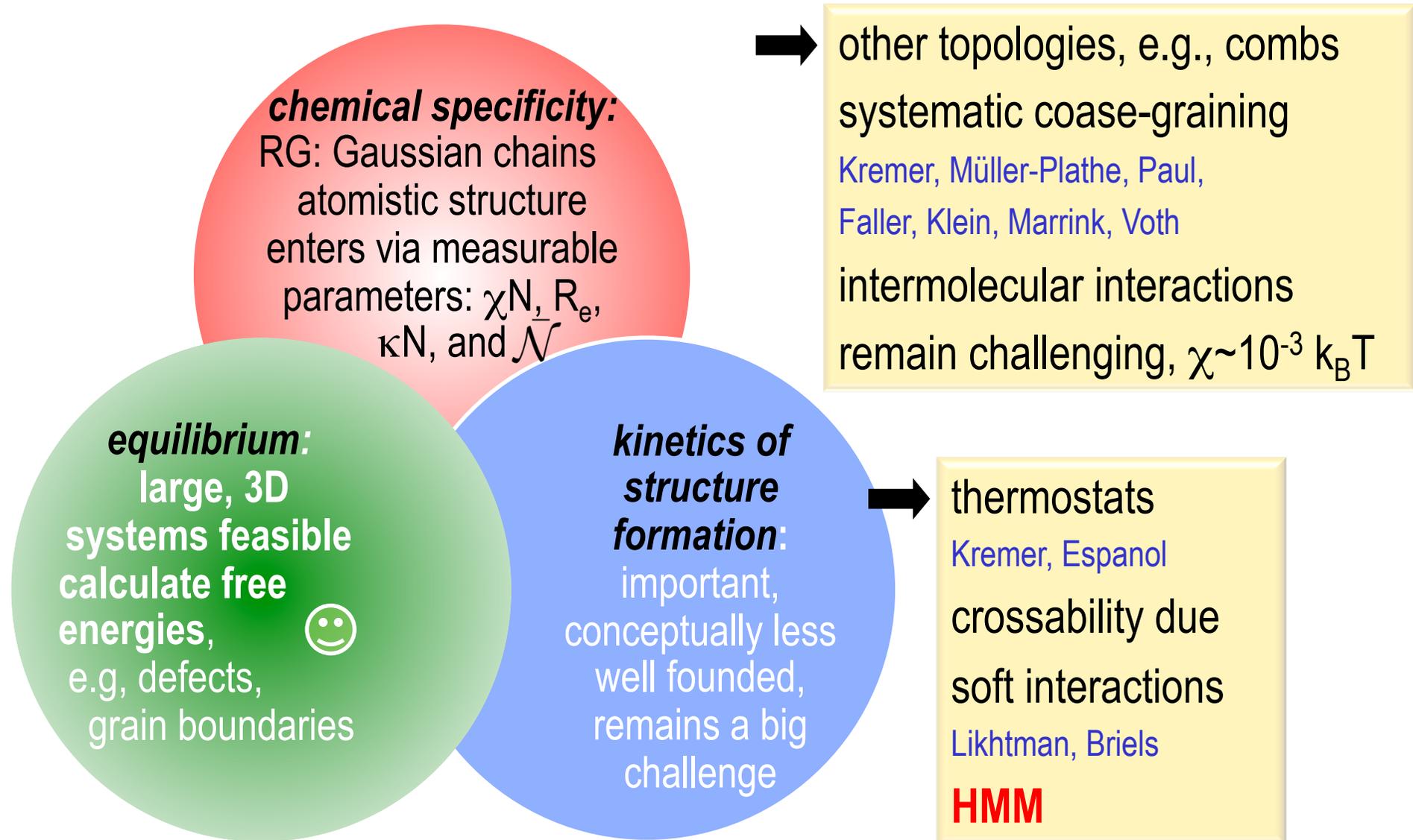
string of morphologies for stalk formation



free-energy barrier for stalk formation



Mesososcopic and Monte Carlo Simulations



thanks to **Kostas Daoulas, Yuliya Smirnova, An-Chang Shi,**
Juan de Pablo / Paul Nealey

Moscow, July 18, 2012