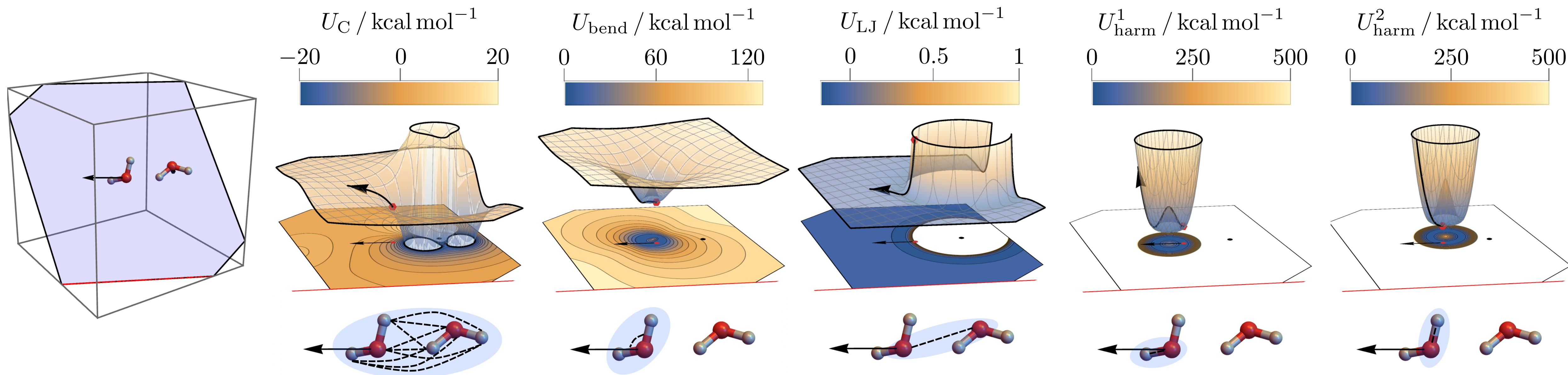


# Bringing the Power of Monte Carlo methods to Long-Range-Interacting Molecular Systems

Philipp Höllmer (AG Monien)

09. November 2022, Journal Club on Condensed Matter Theory, University of Bonn

Collaborators: N. Noirault (ENS Paris), B. Li (ENS Paris), A. C. Maggs (ESPCI Paris), W. Krauth (ENS Paris)

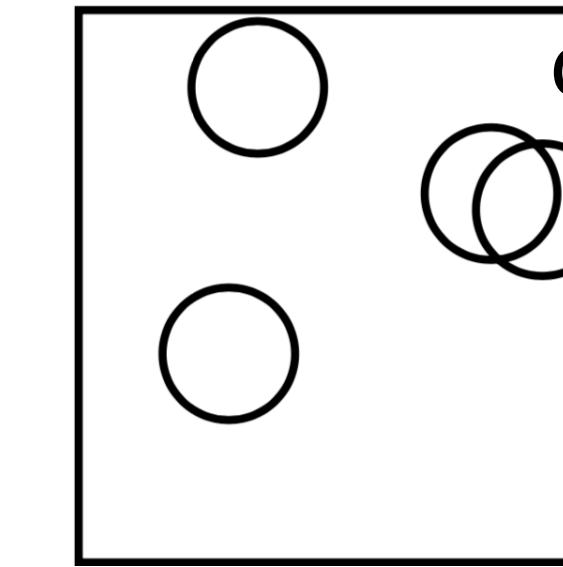
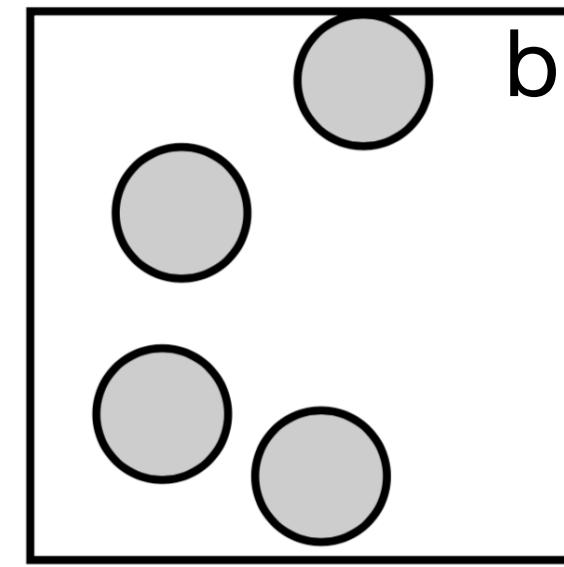
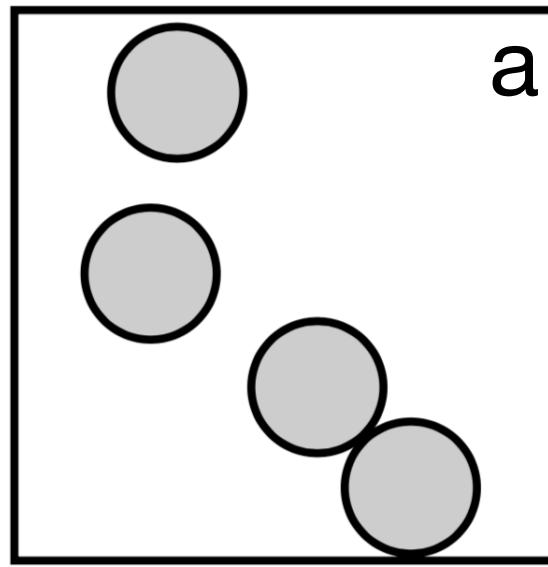


# Structure

1. Non-Reversible Event-Chain Monte Carlo (ECMC) for the Hard-Disk Model.
2. Generalization of ECMC to Molecular Systems.
3. ECMC Variants and Simulation Results.

# Hard-Disk Model

- Probability density of configuration  $c = \{\mathbf{r}_1, \mathbf{r}_2, \dots, \mathbf{r}_N\}$ :  $\pi(c) \propto \begin{cases} 1 & \text{if } c \text{ legal,} \\ 0 & \text{otherwise.} \end{cases}$



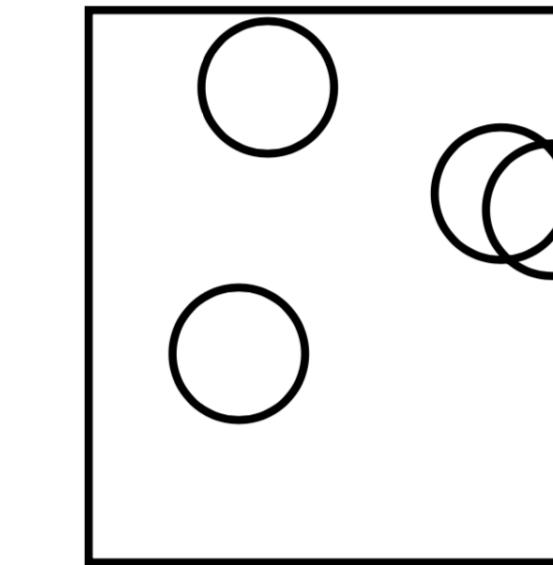
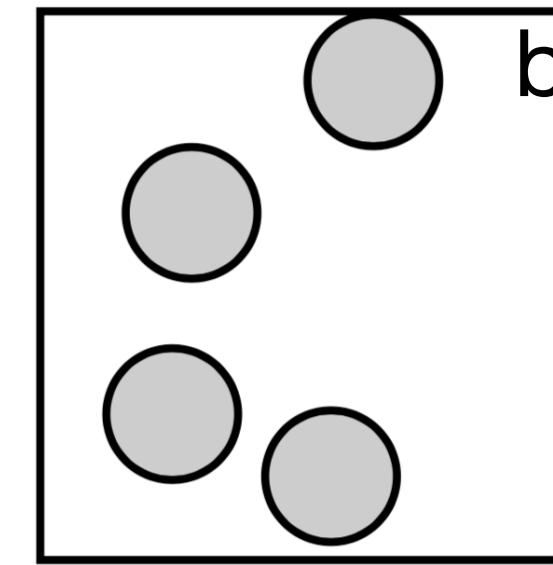
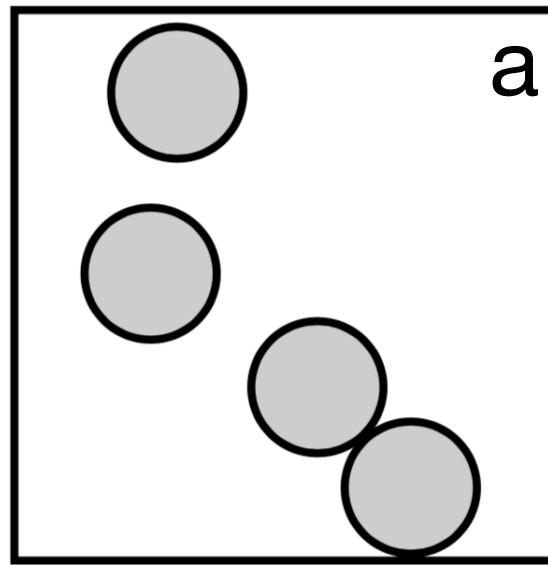
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W. Krauth, *Statistical Mechanics: Algorithms and Computations* (Oxford University Press, 2006)

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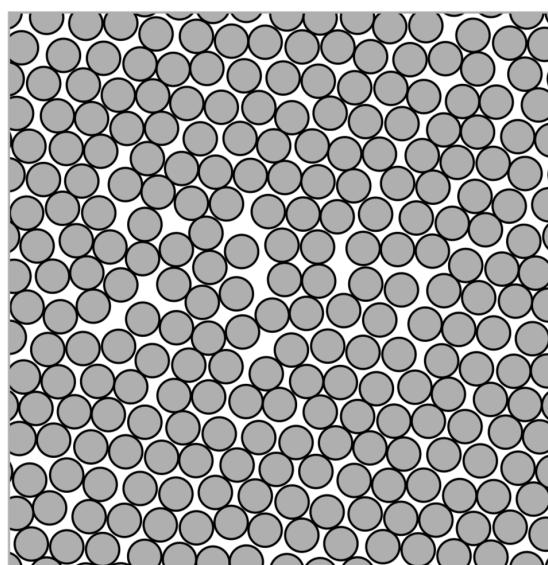


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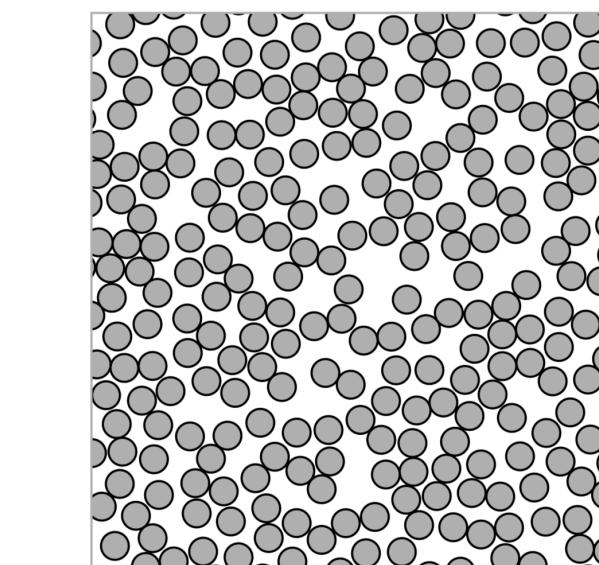
- Hard-disk model has a phase transition when hard-disk density is changed:



Solid



Precise melting scenario?



Fluid

E. Bernard, *Algorithms and applications of the Monte Carlo method: Two-dimensional melting and perfect sampling* (PhD thesis, 2011)

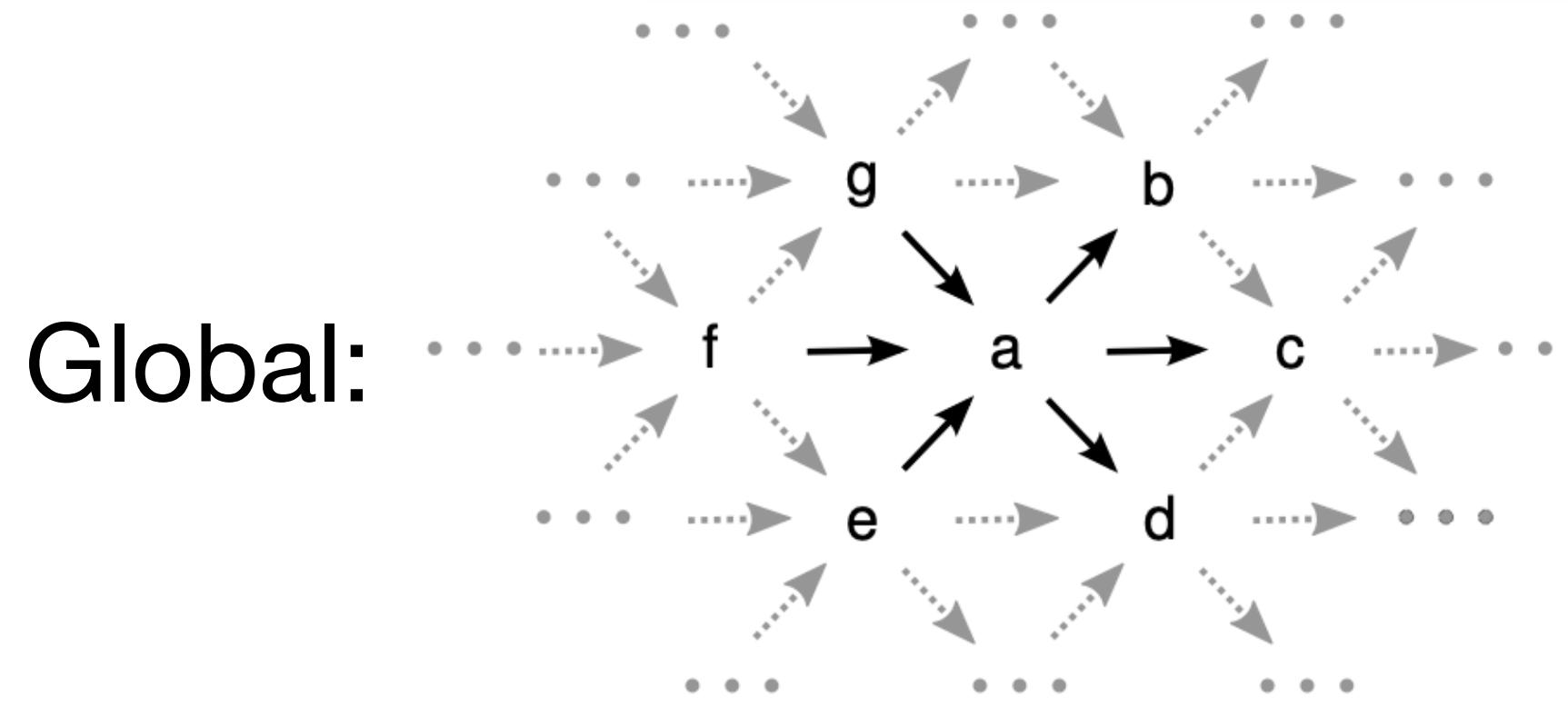
# Markov-Chain Monte Carlo (MCMC)

- Idea: Change configuration from  $c$  to  $c'$  with transition probability  $p(c \rightarrow c')$ .  
→ Only converges to probability distribution  $\pi$  if:

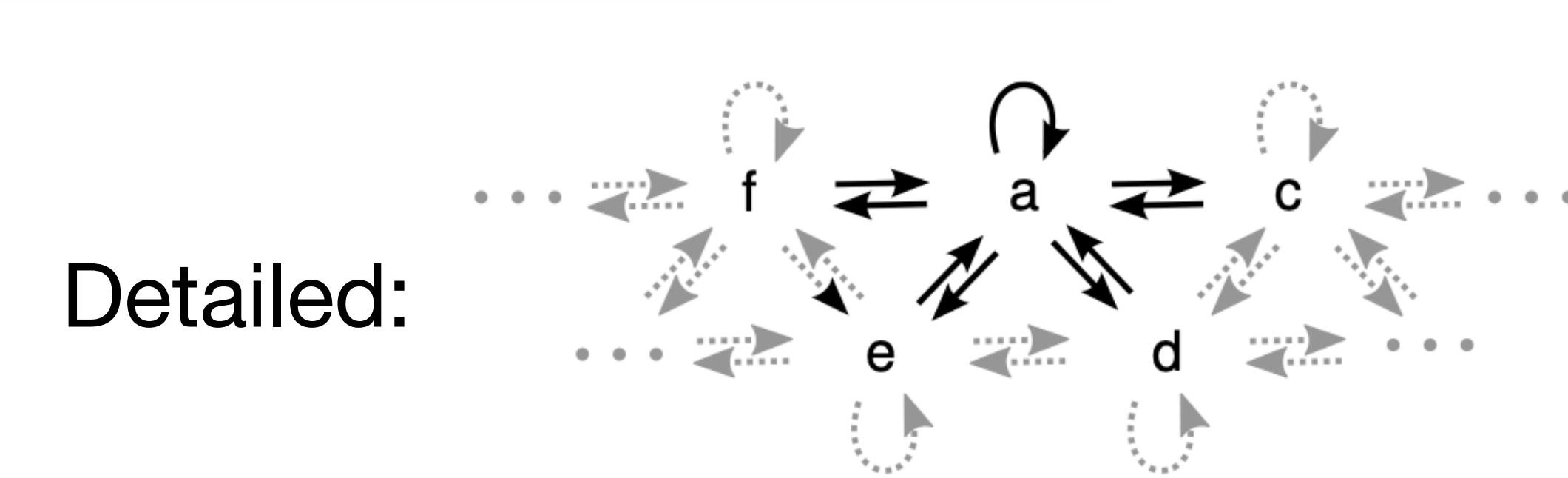
$$\sum_{c'} \pi(c') p(c' \rightarrow c) = \pi(c) \quad (\text{global balance})$$

- Reversible** Markov chains obey global balance by satisfying a weaker condition:

$$\pi(c') p(c' \rightarrow c) = \pi(c) p(c \rightarrow c') \quad (\text{detailed balance})$$



Michel et al., J. Chem. Phys. 140, 054116 (2014)



Michel et al., J. Chem. Phys. 140, 054116 (2014)

# Reversible Metropolis Algorithm

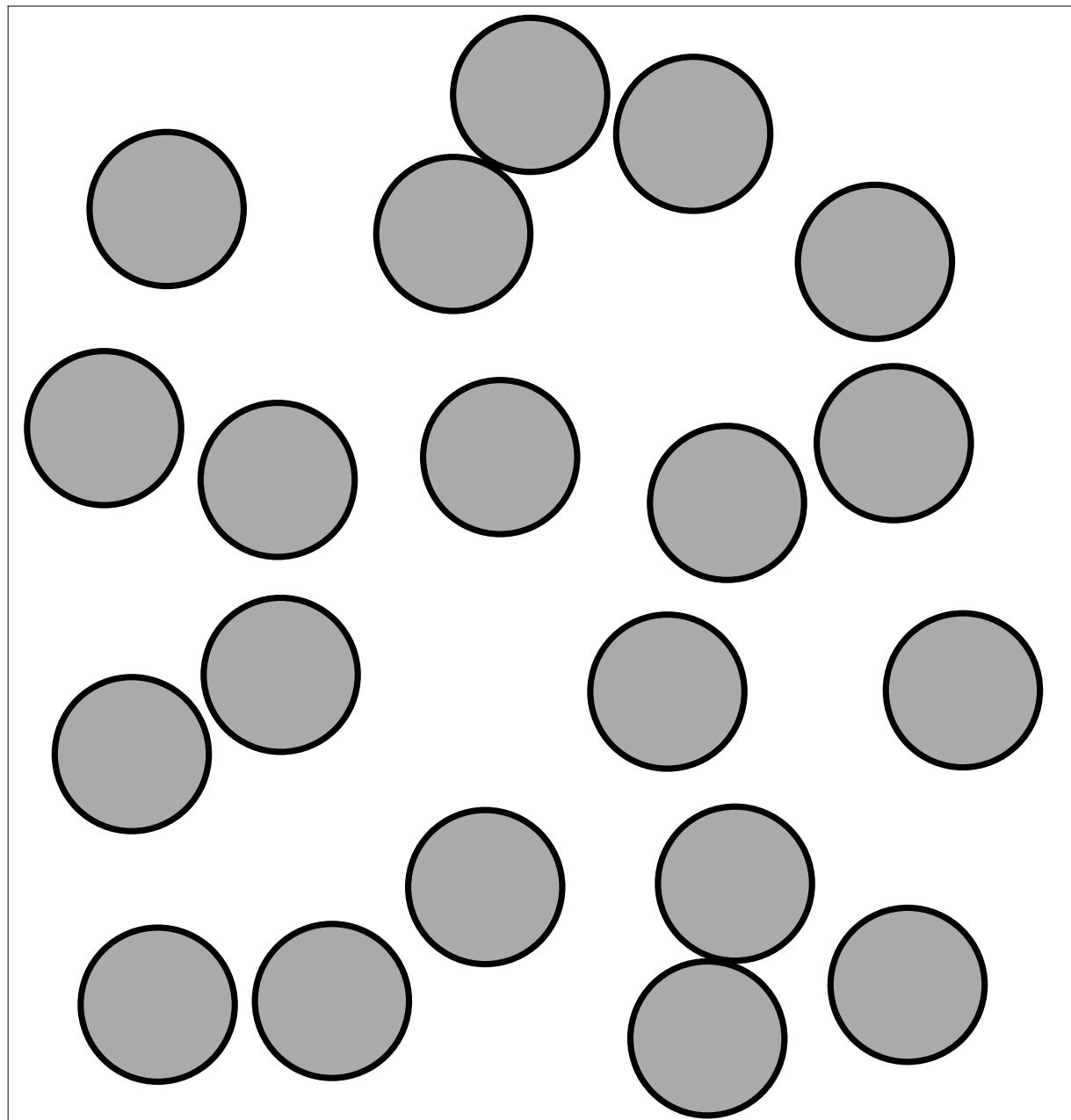
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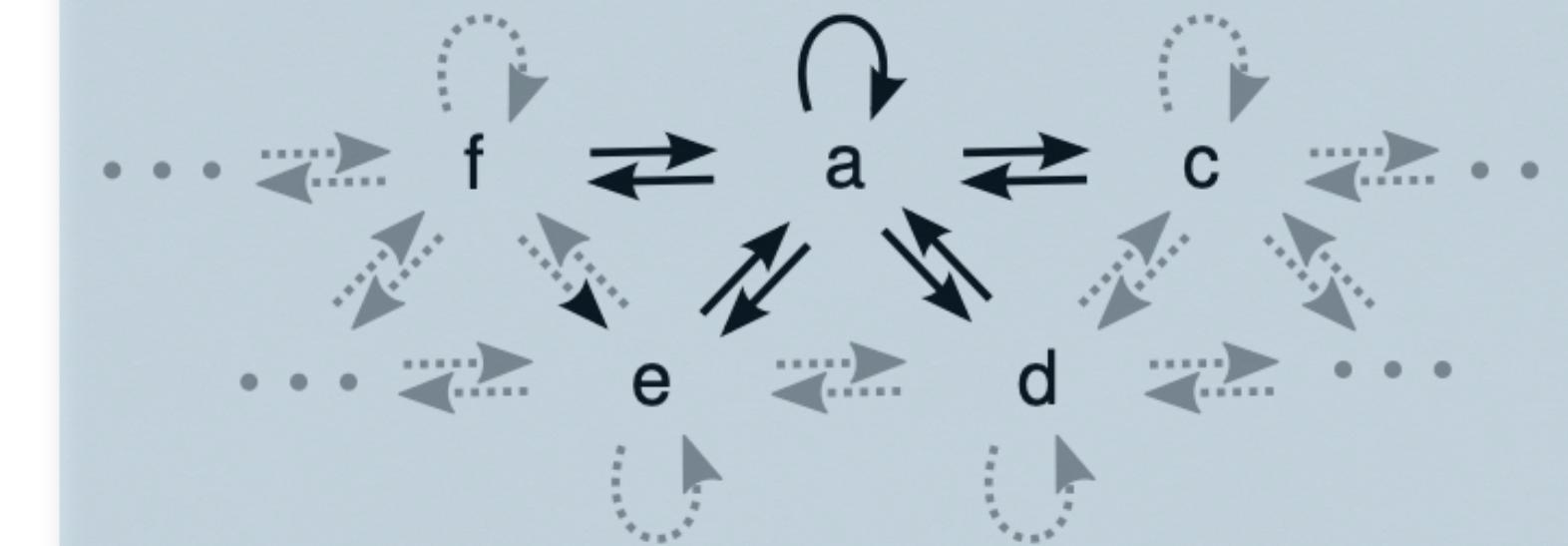
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Obeys detailed balance.



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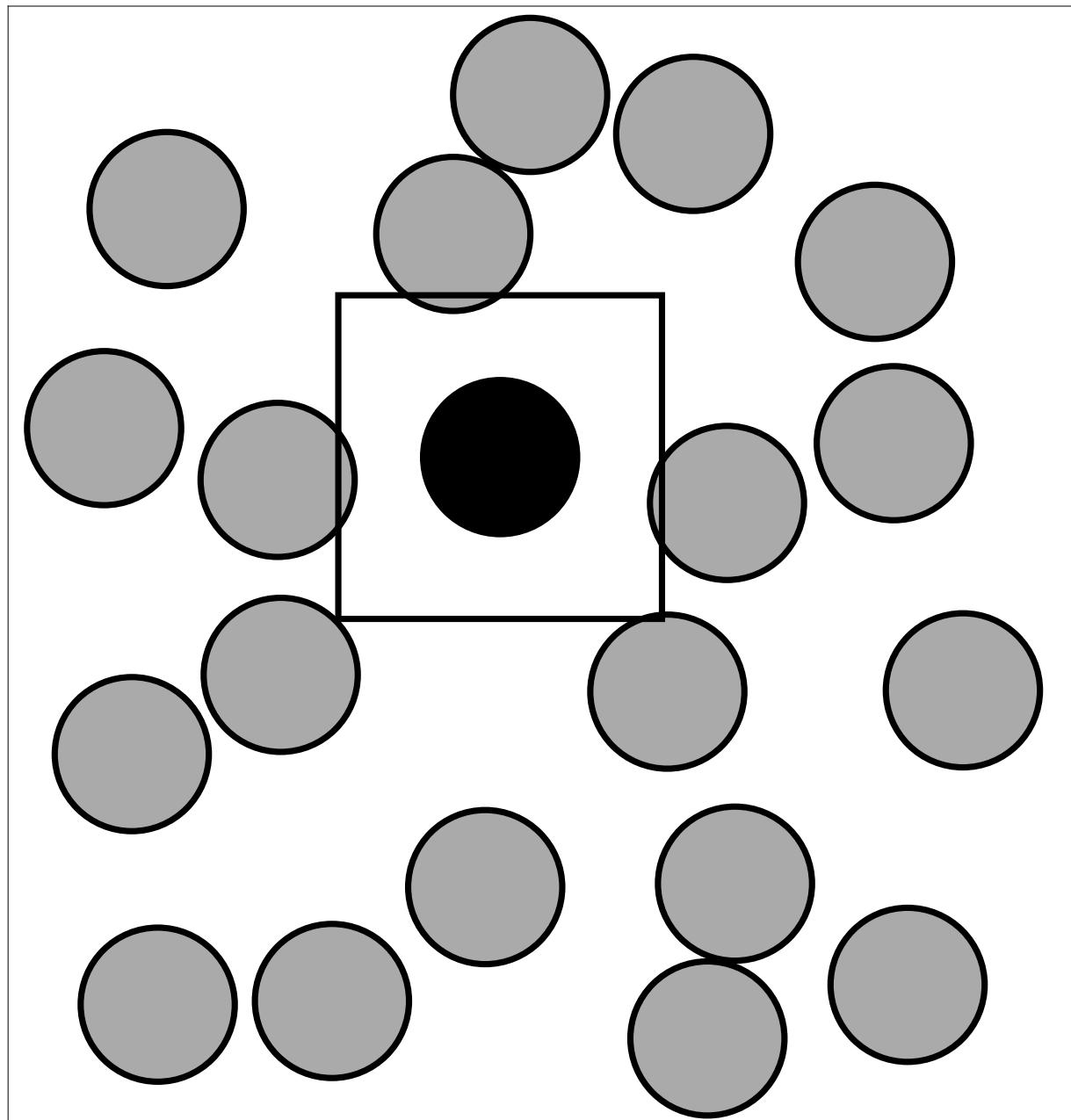
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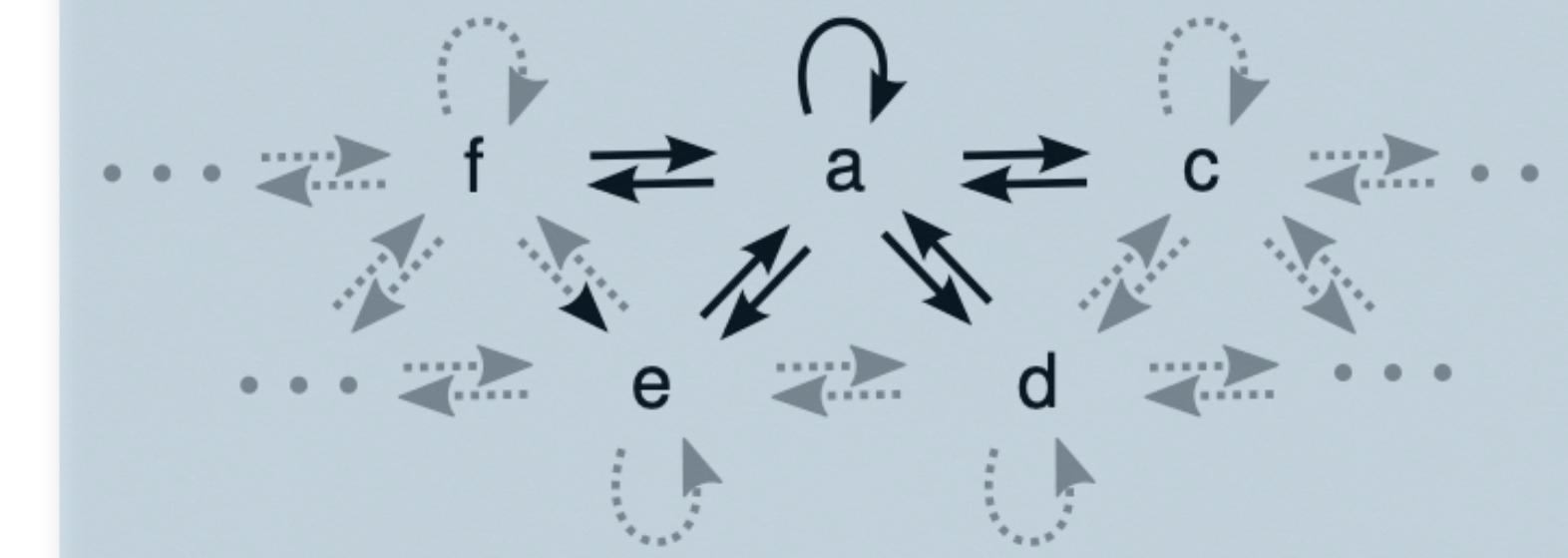
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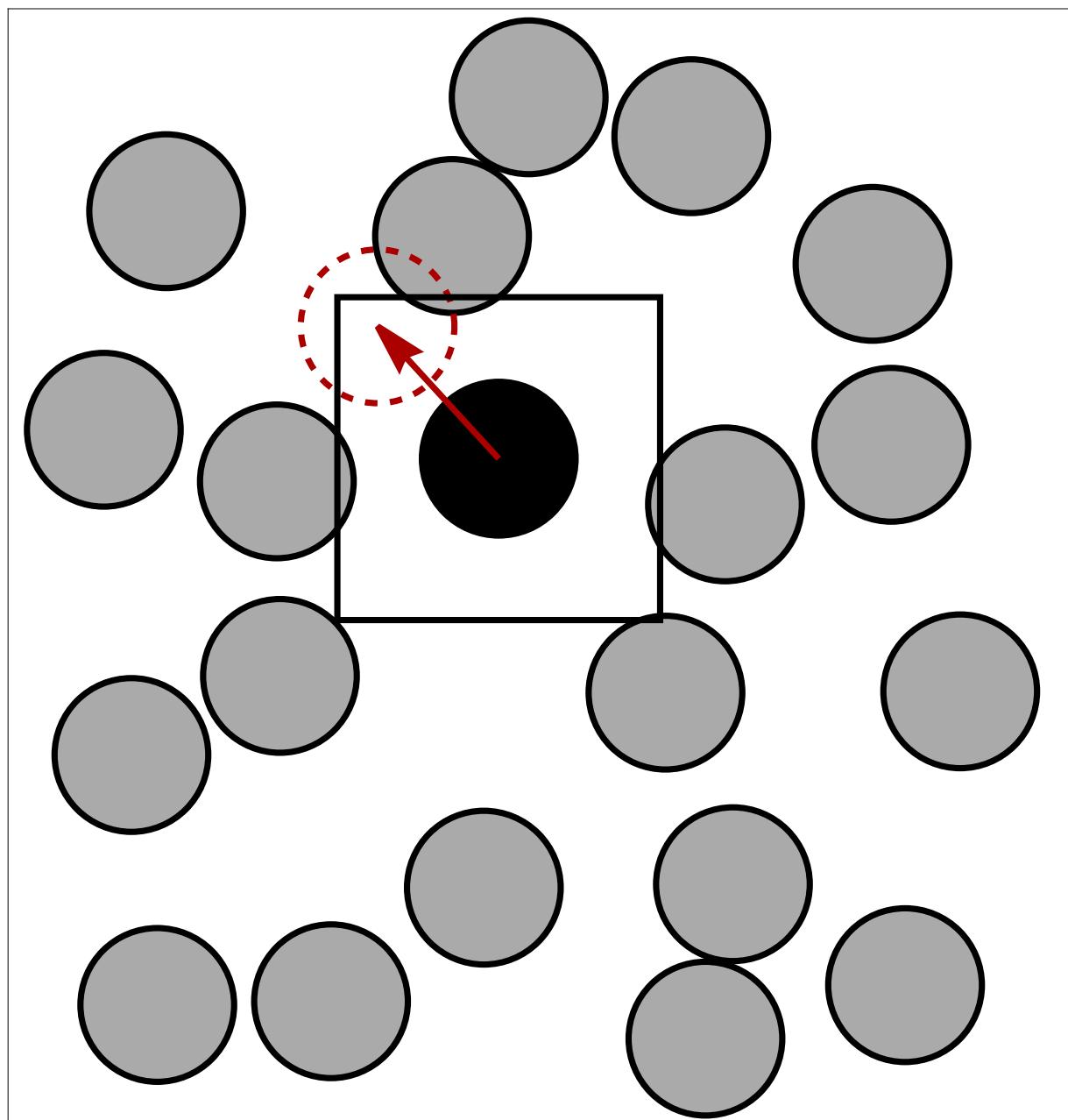
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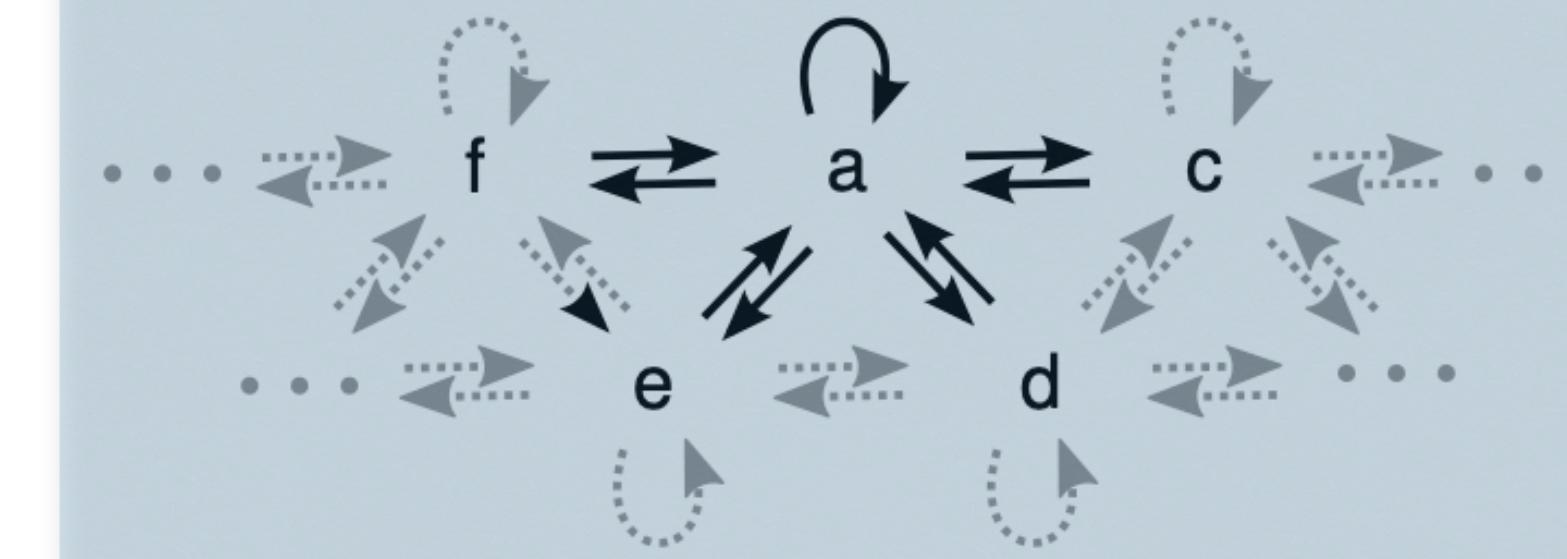
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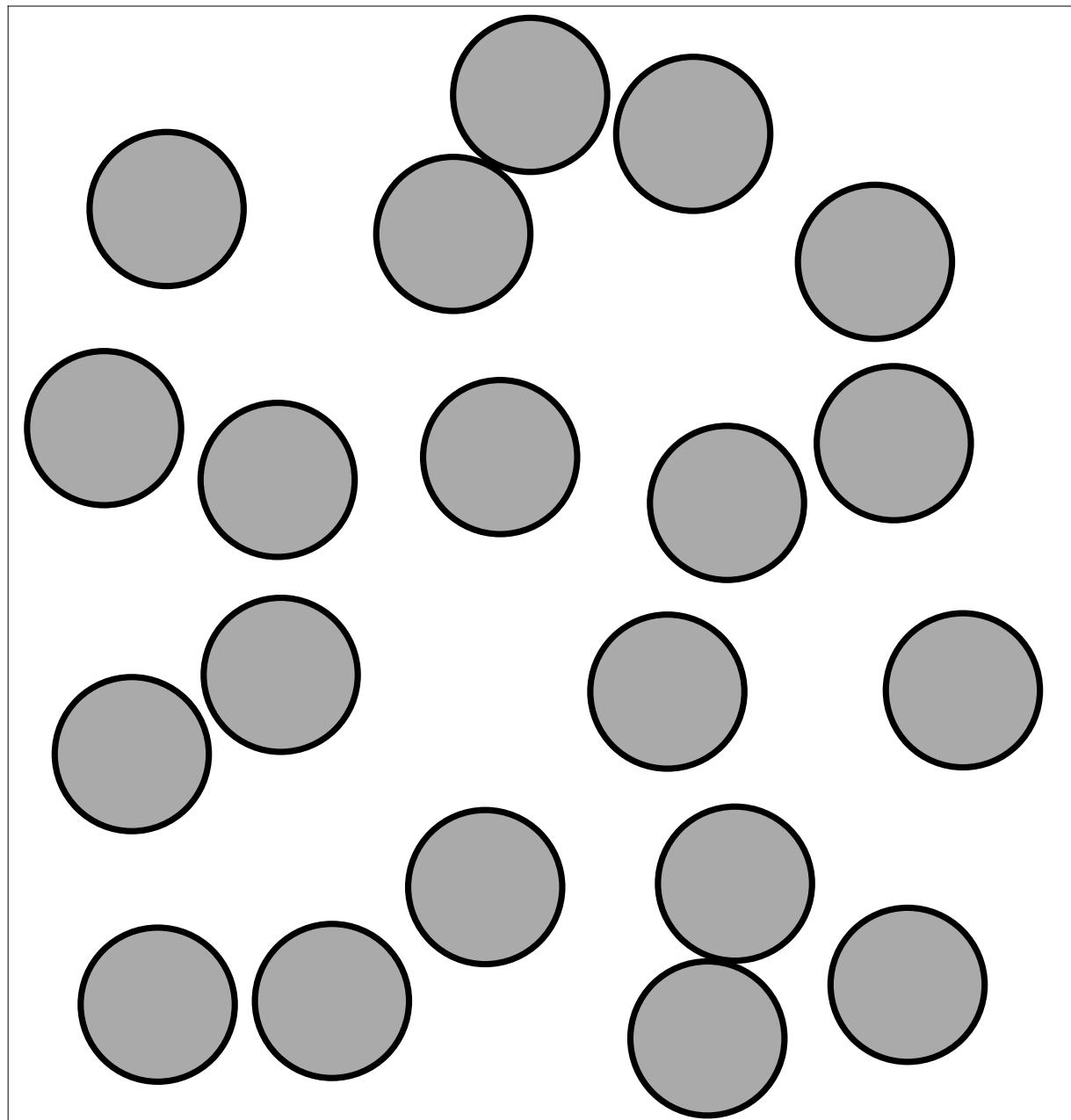
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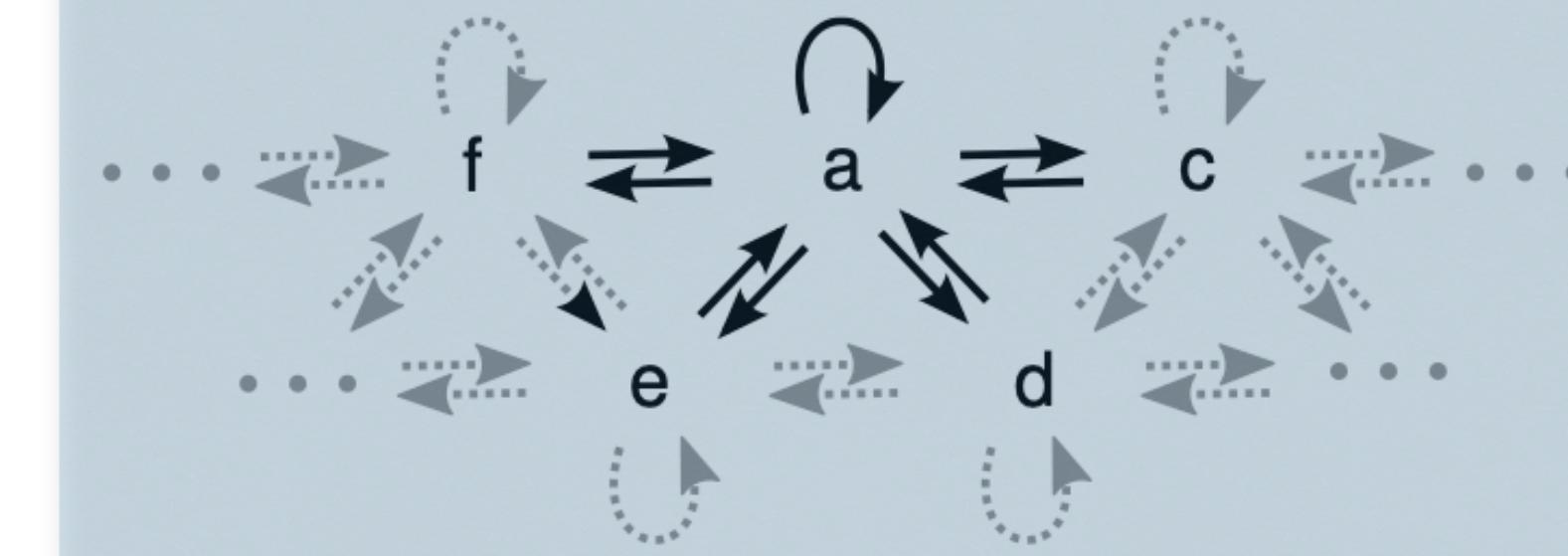
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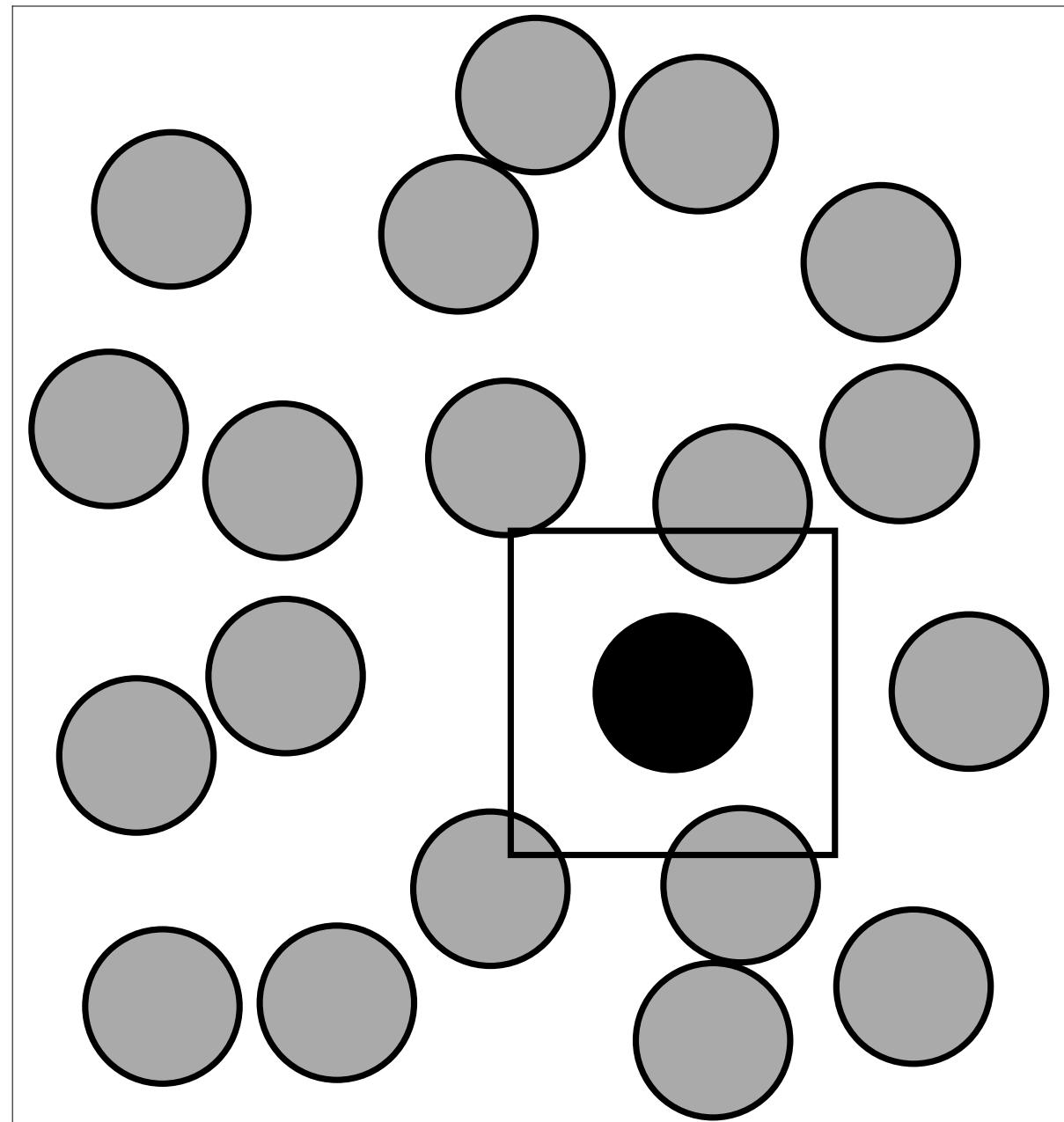
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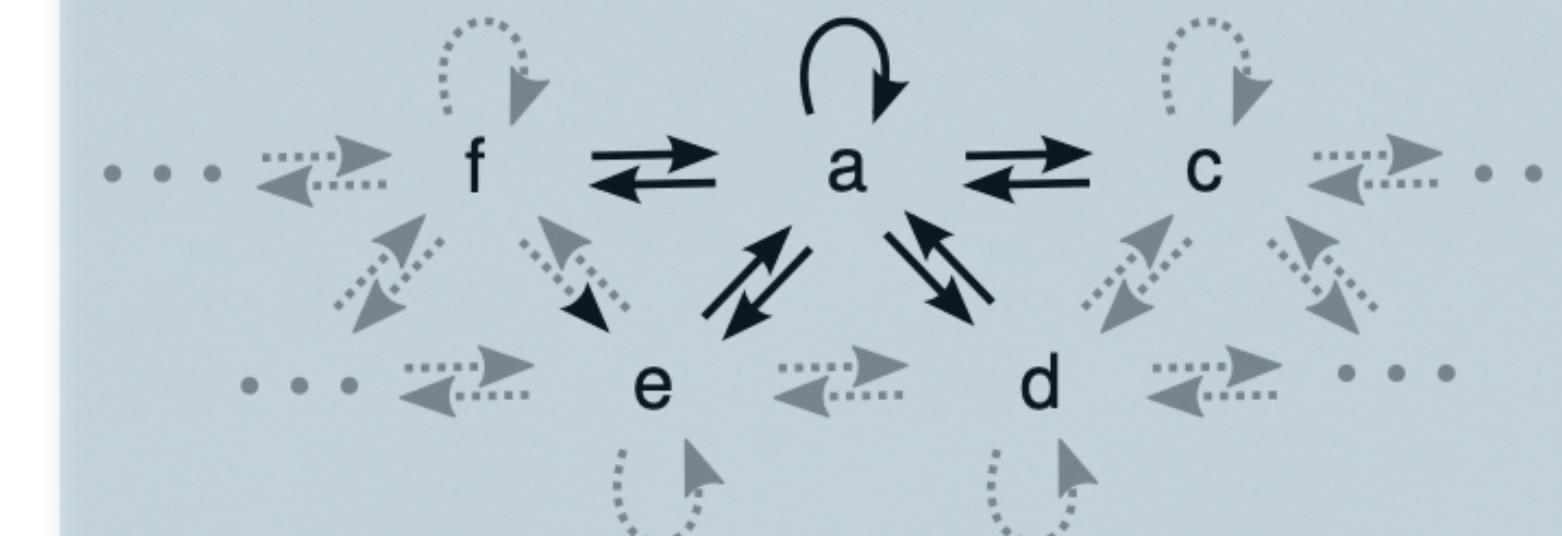
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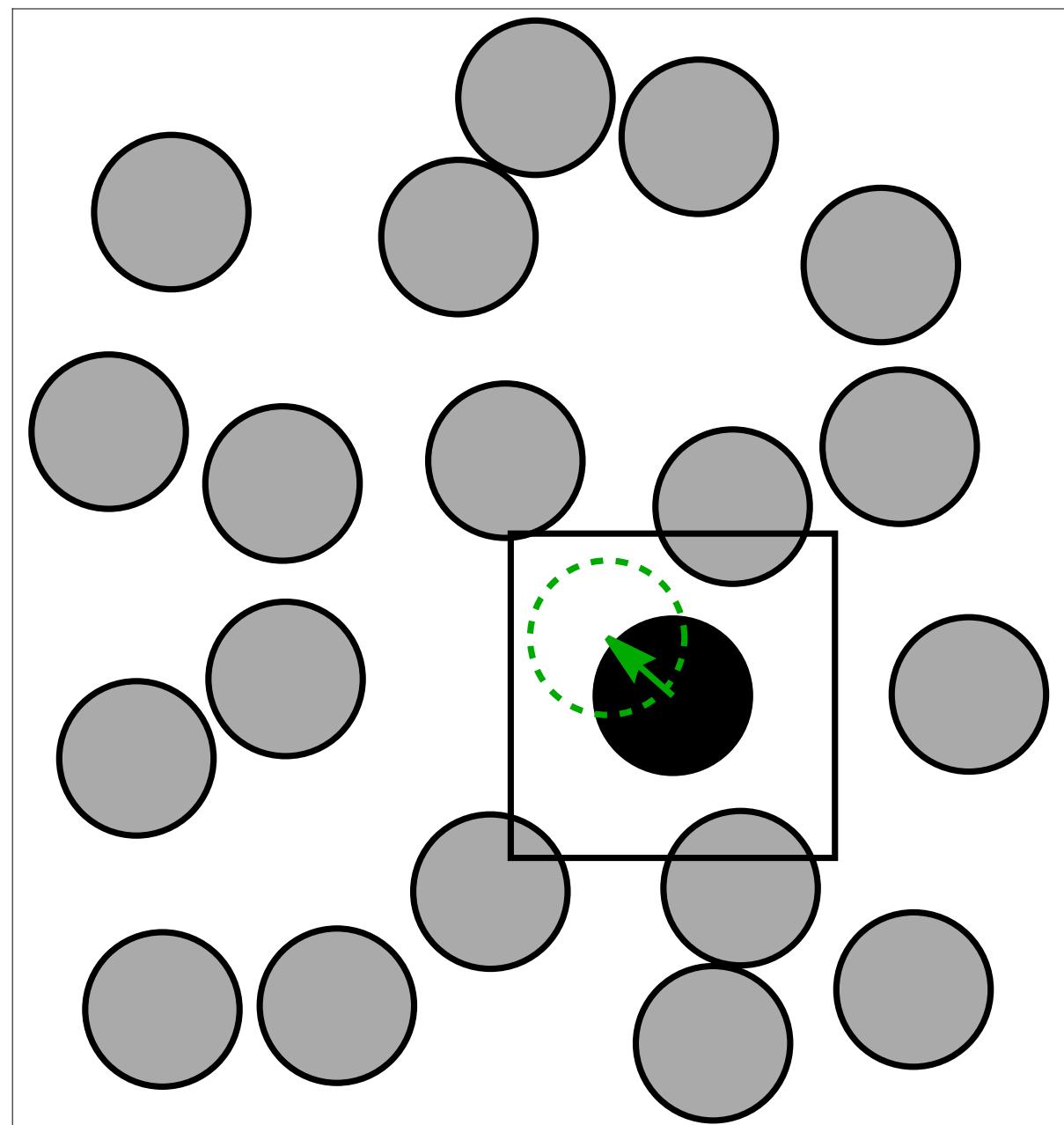
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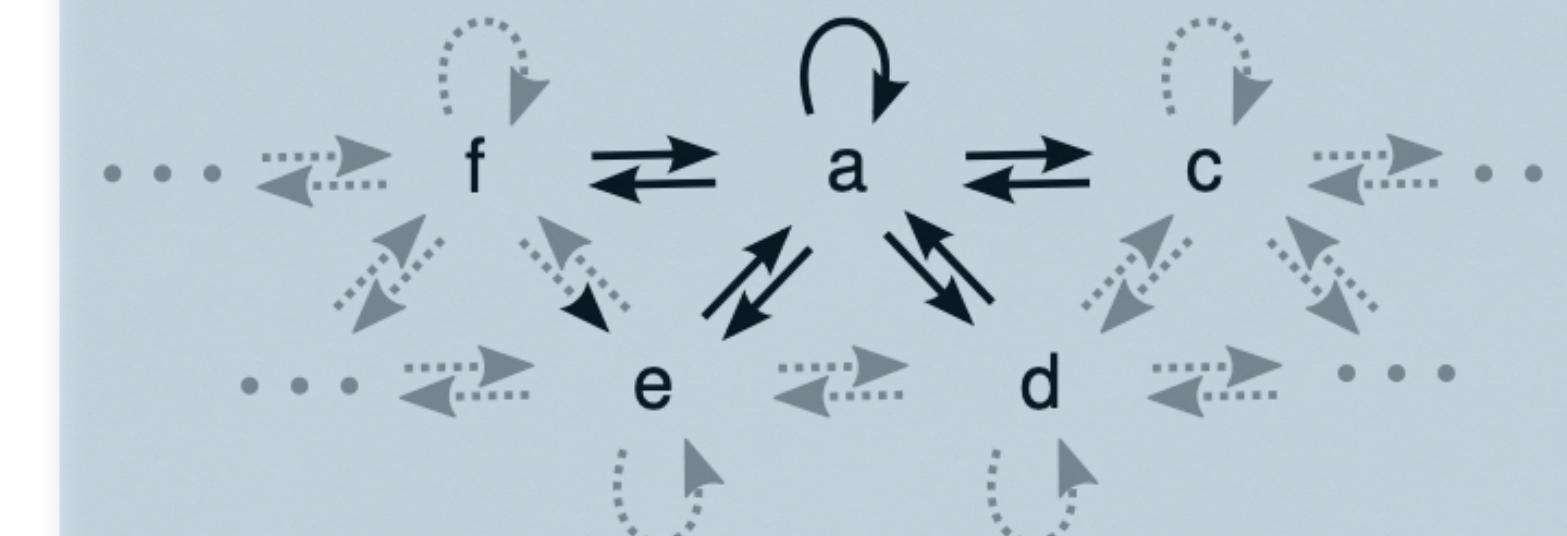
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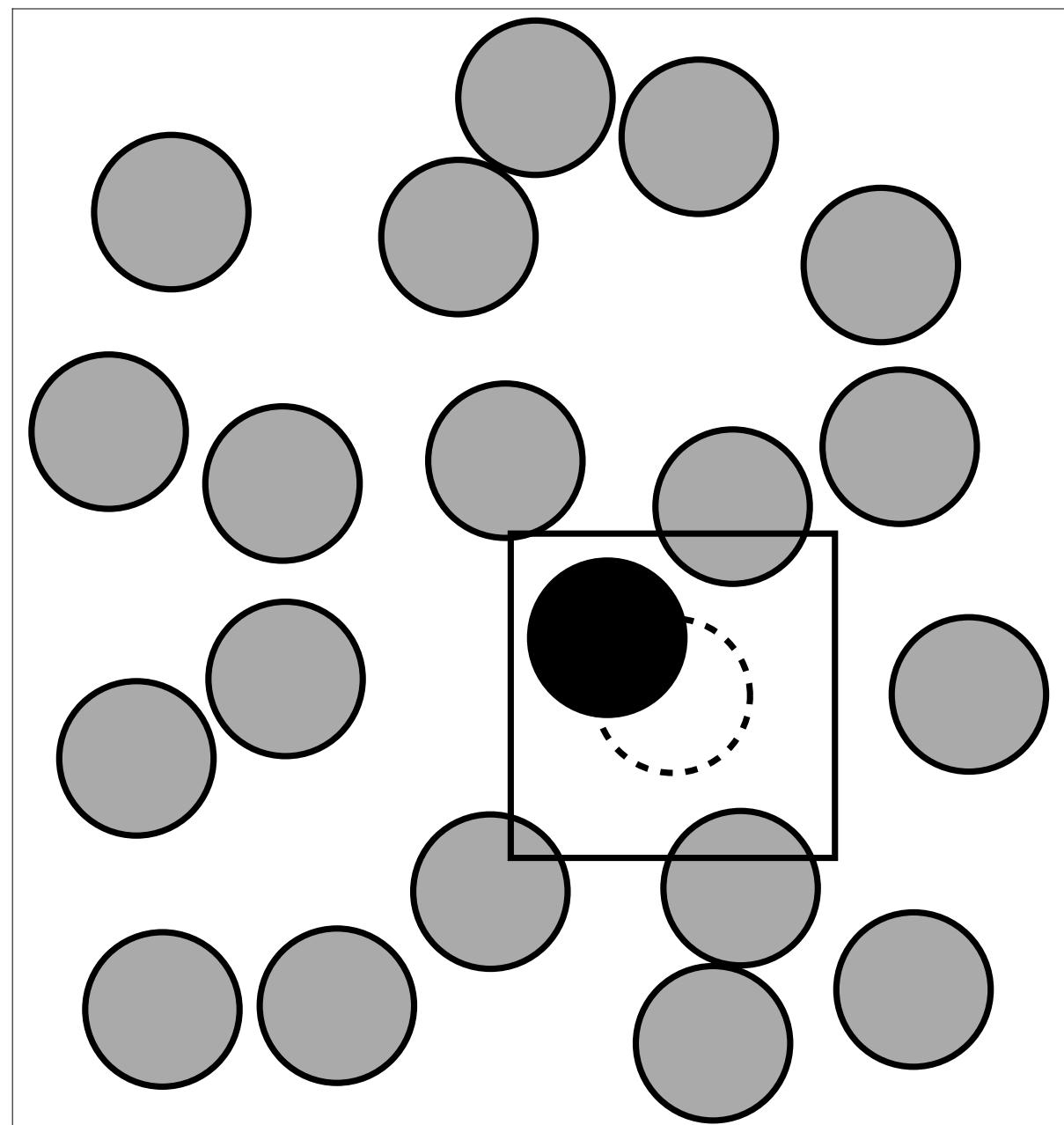
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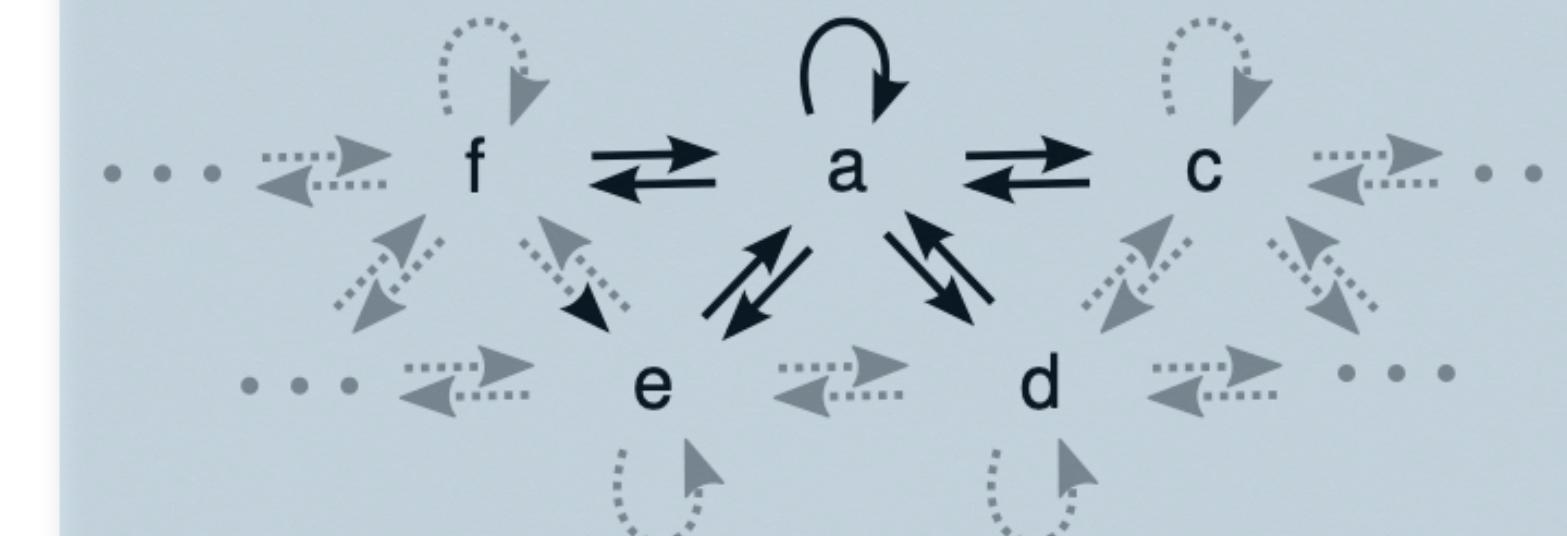
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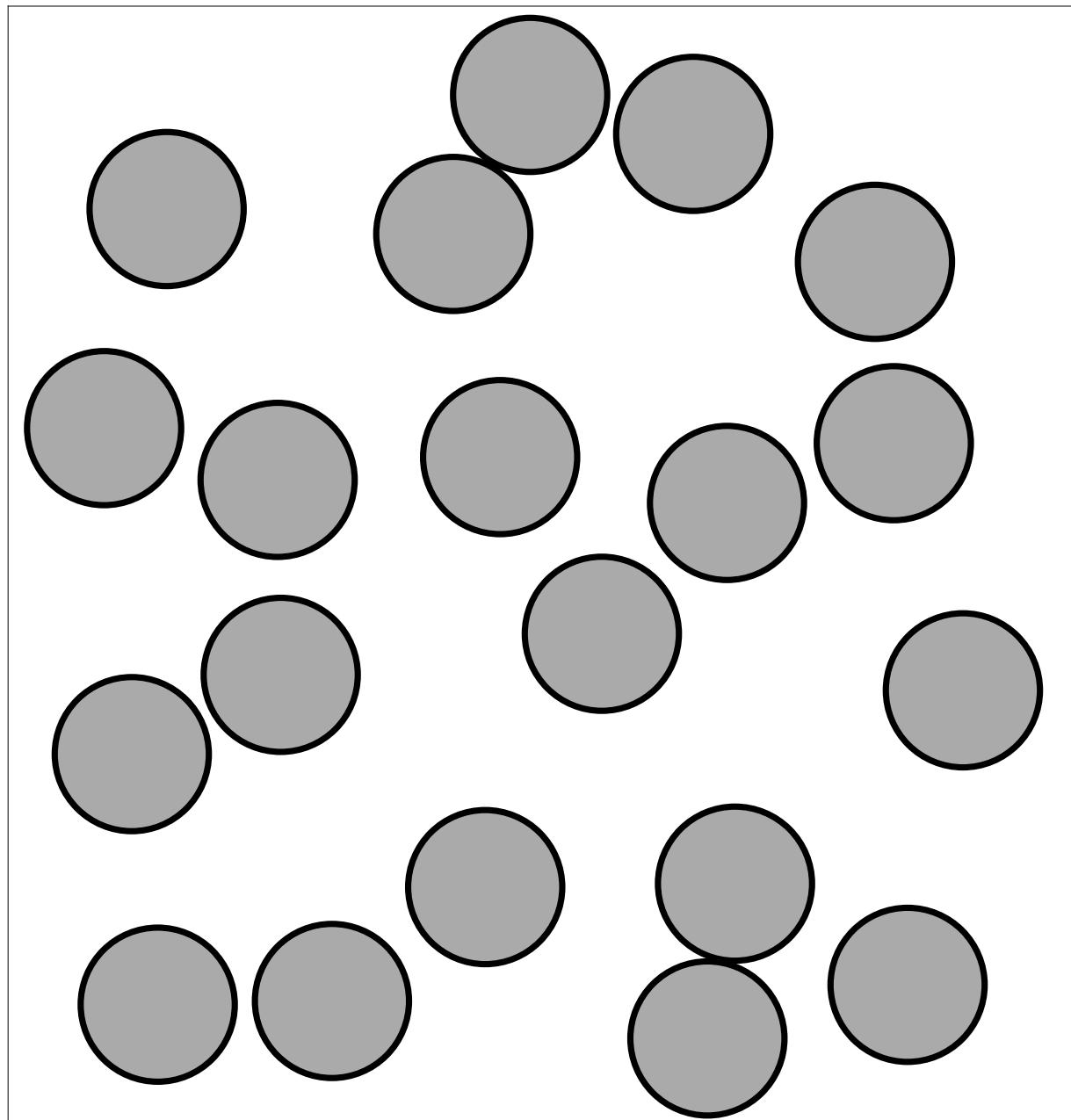
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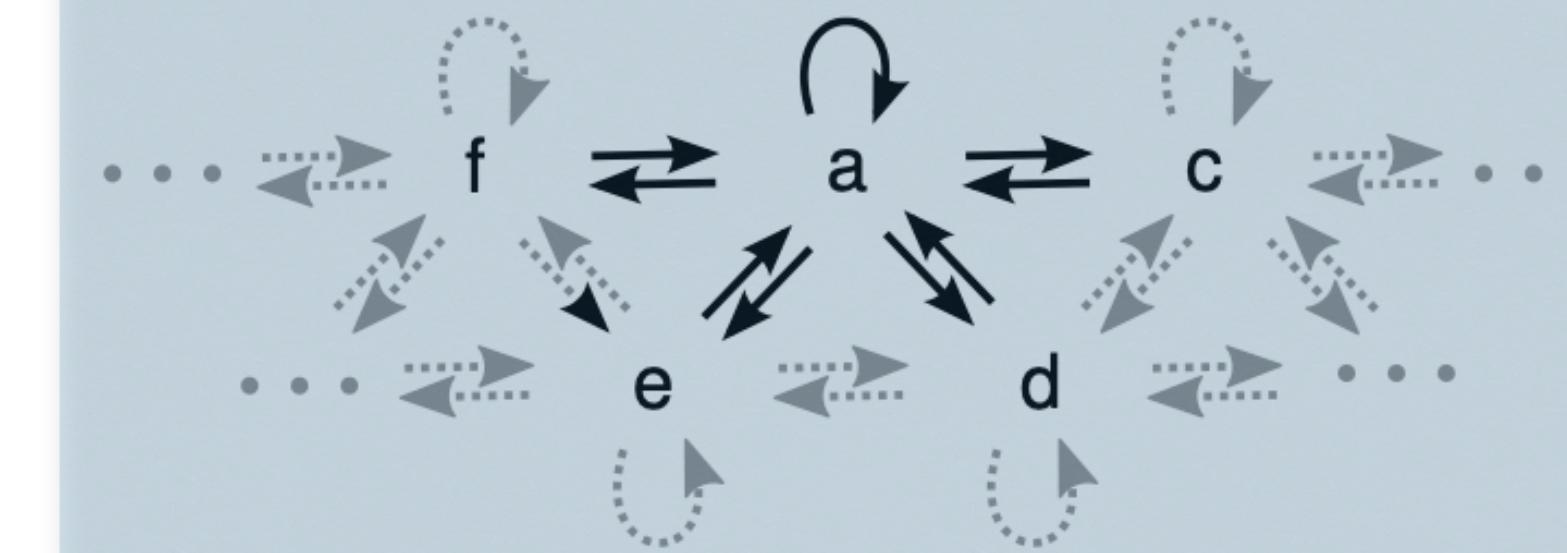
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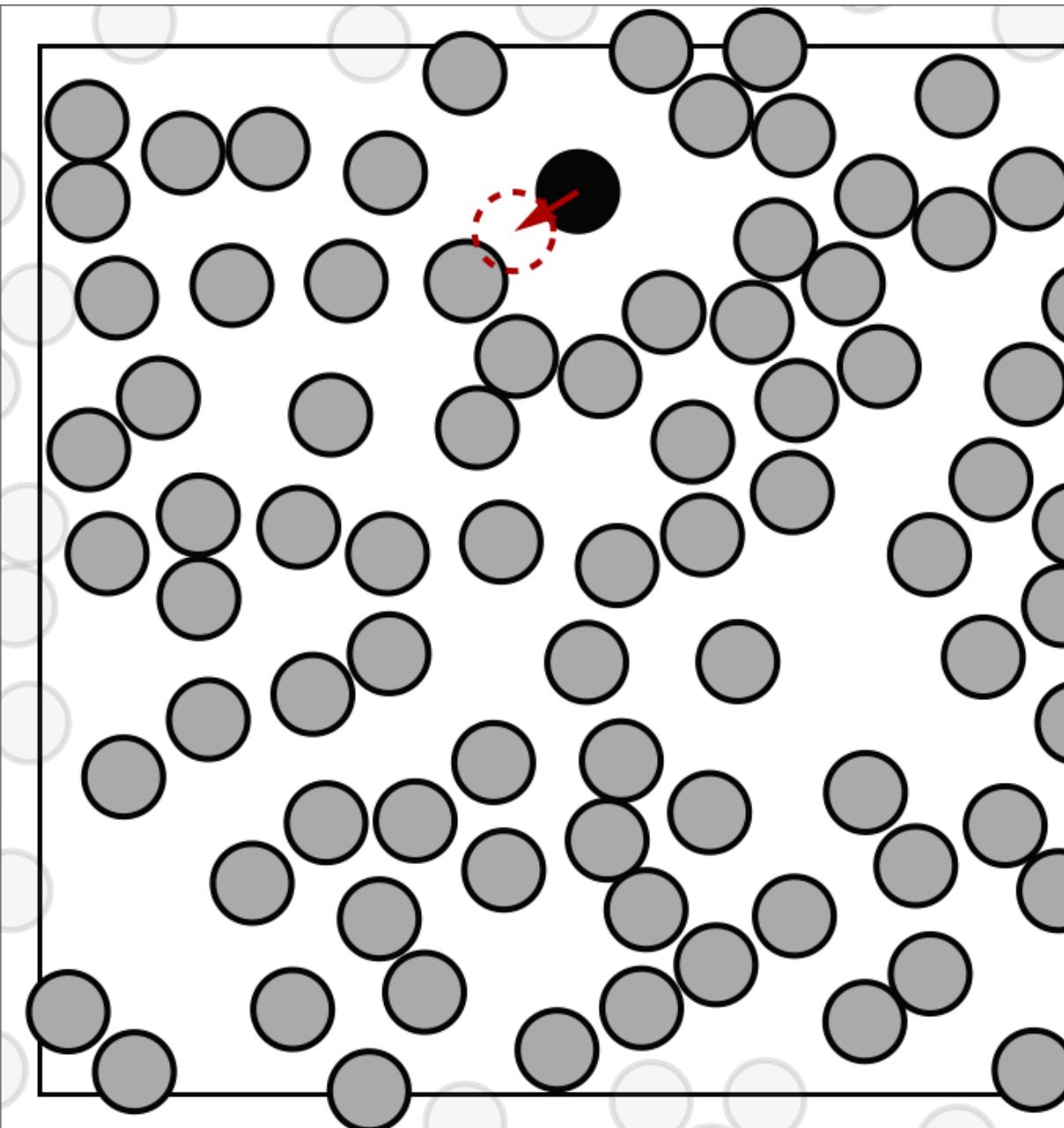
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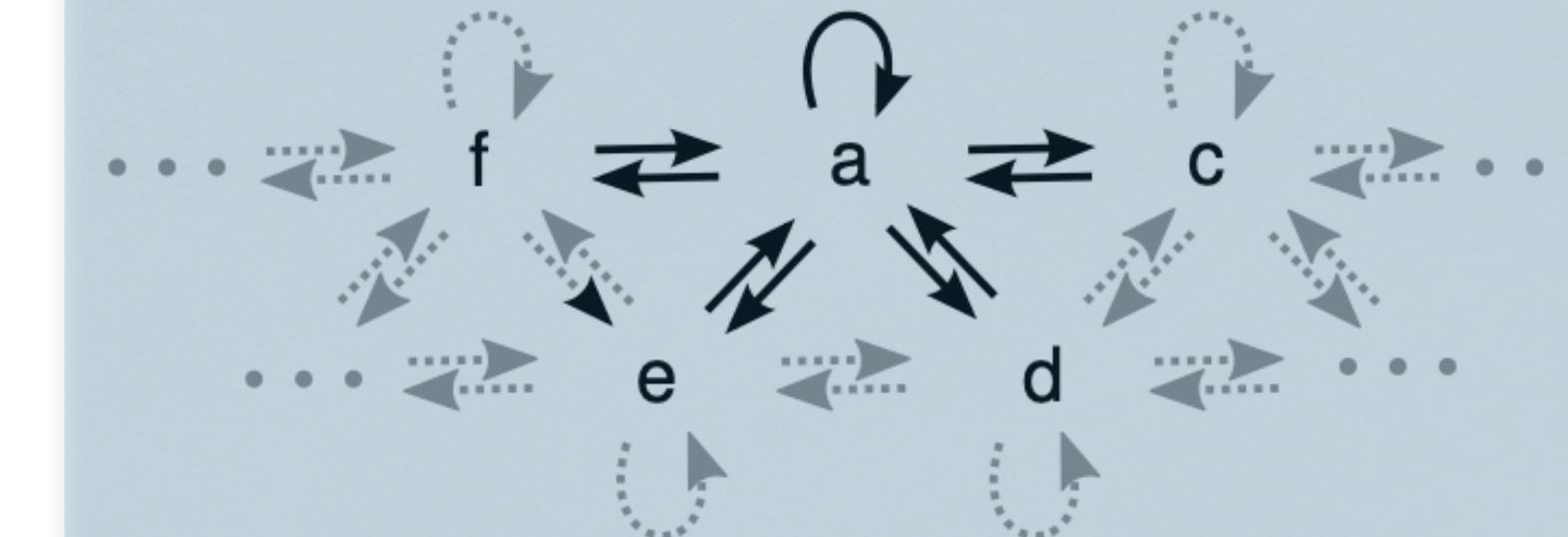
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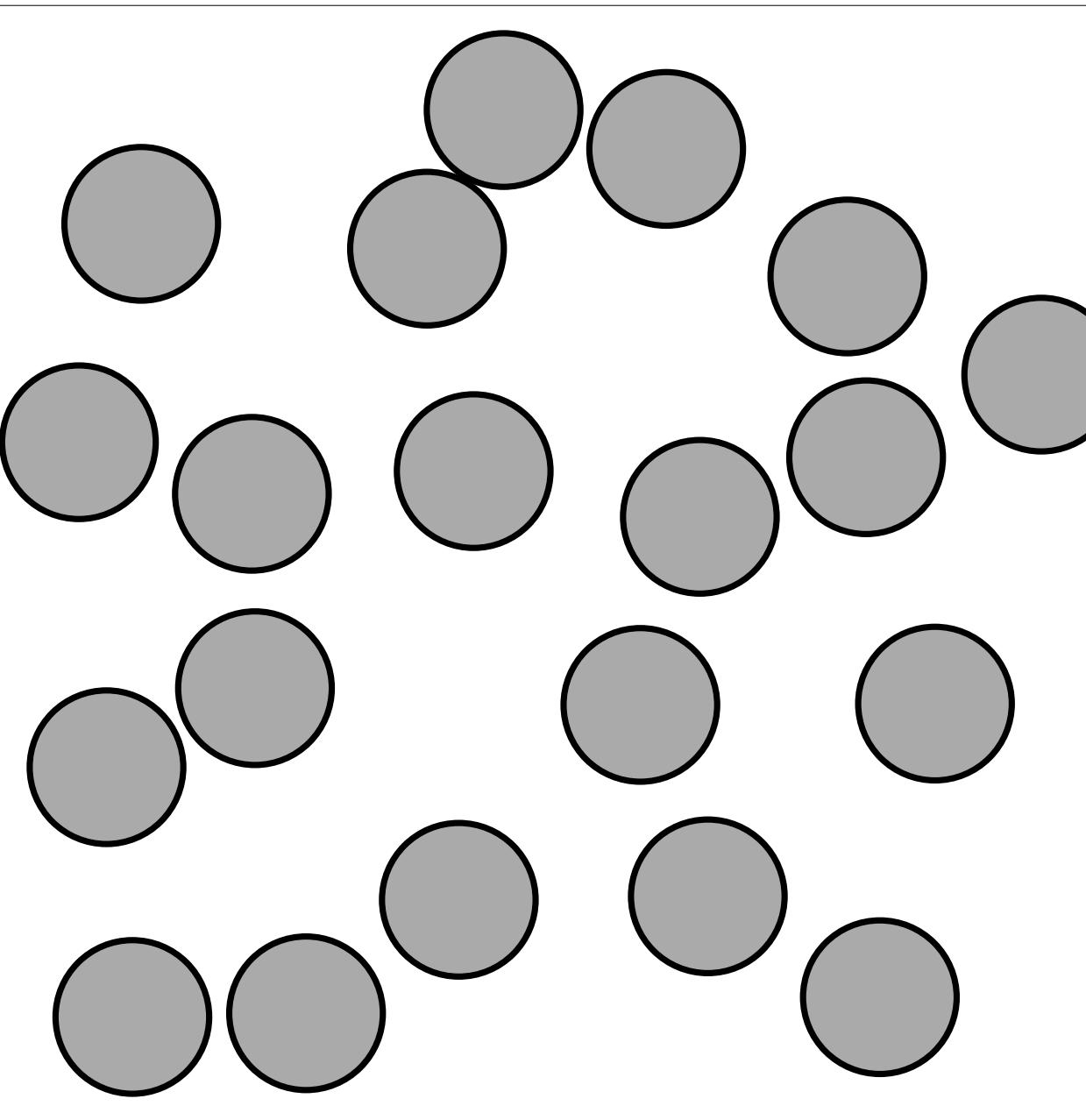
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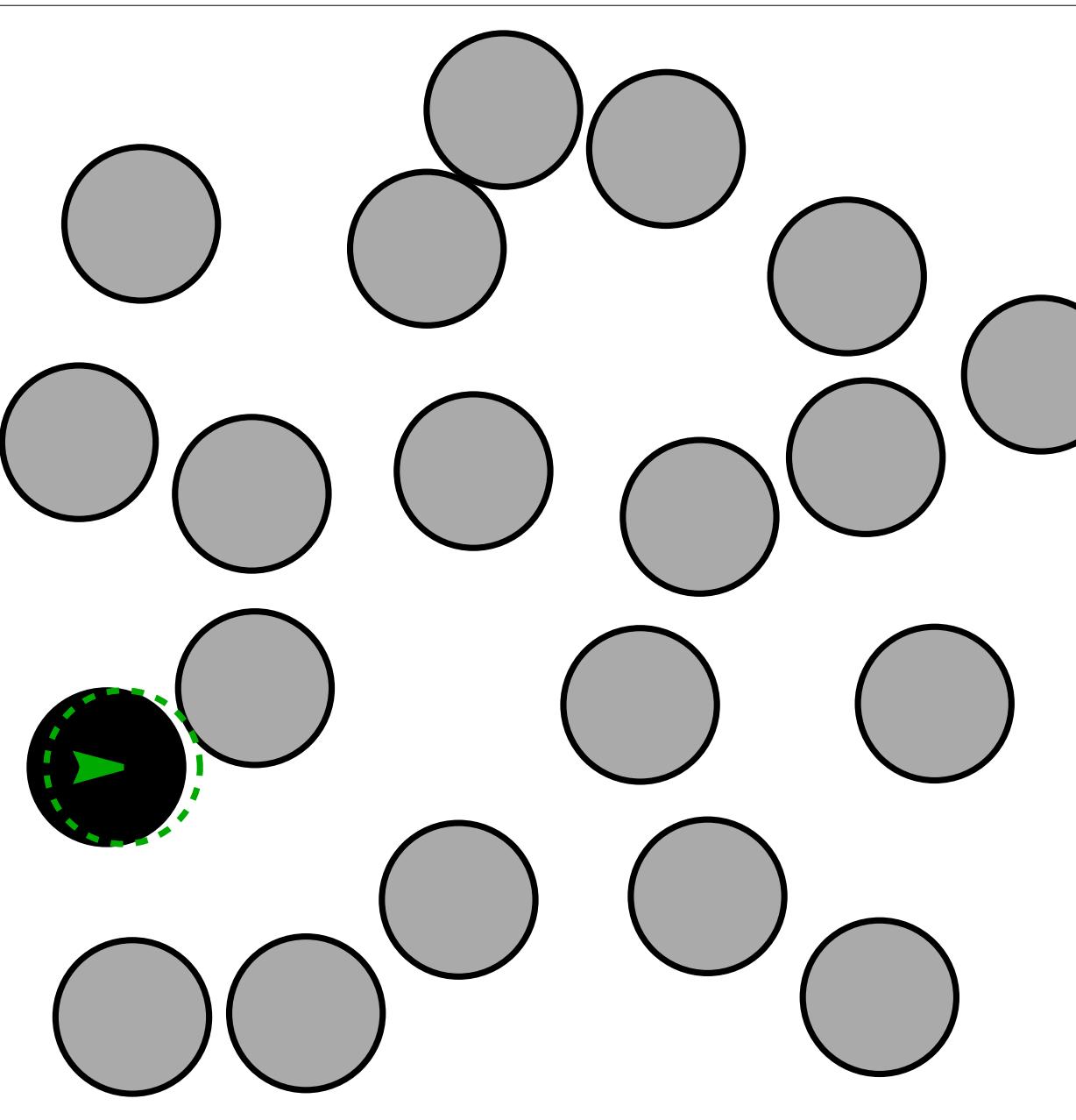
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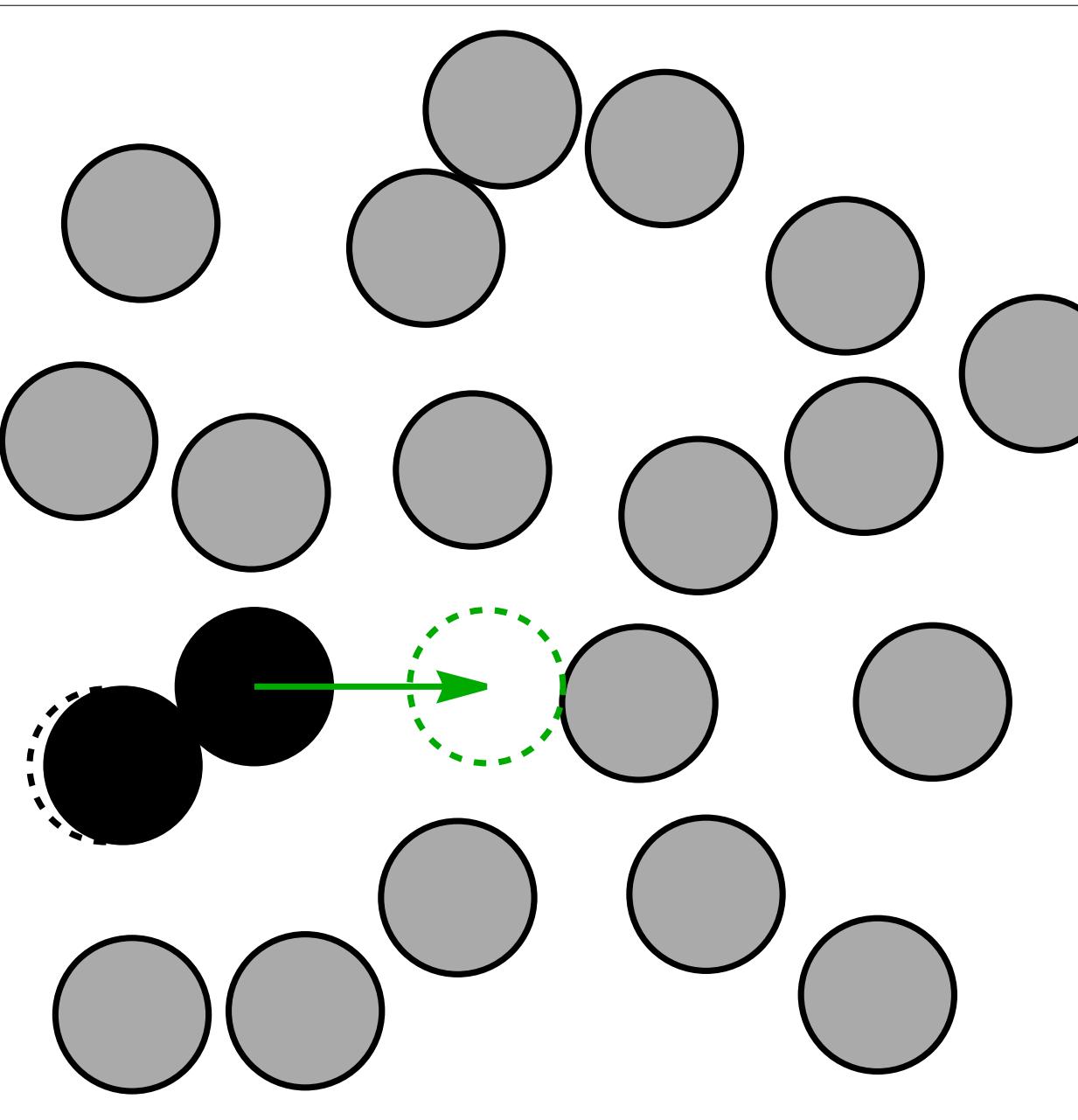
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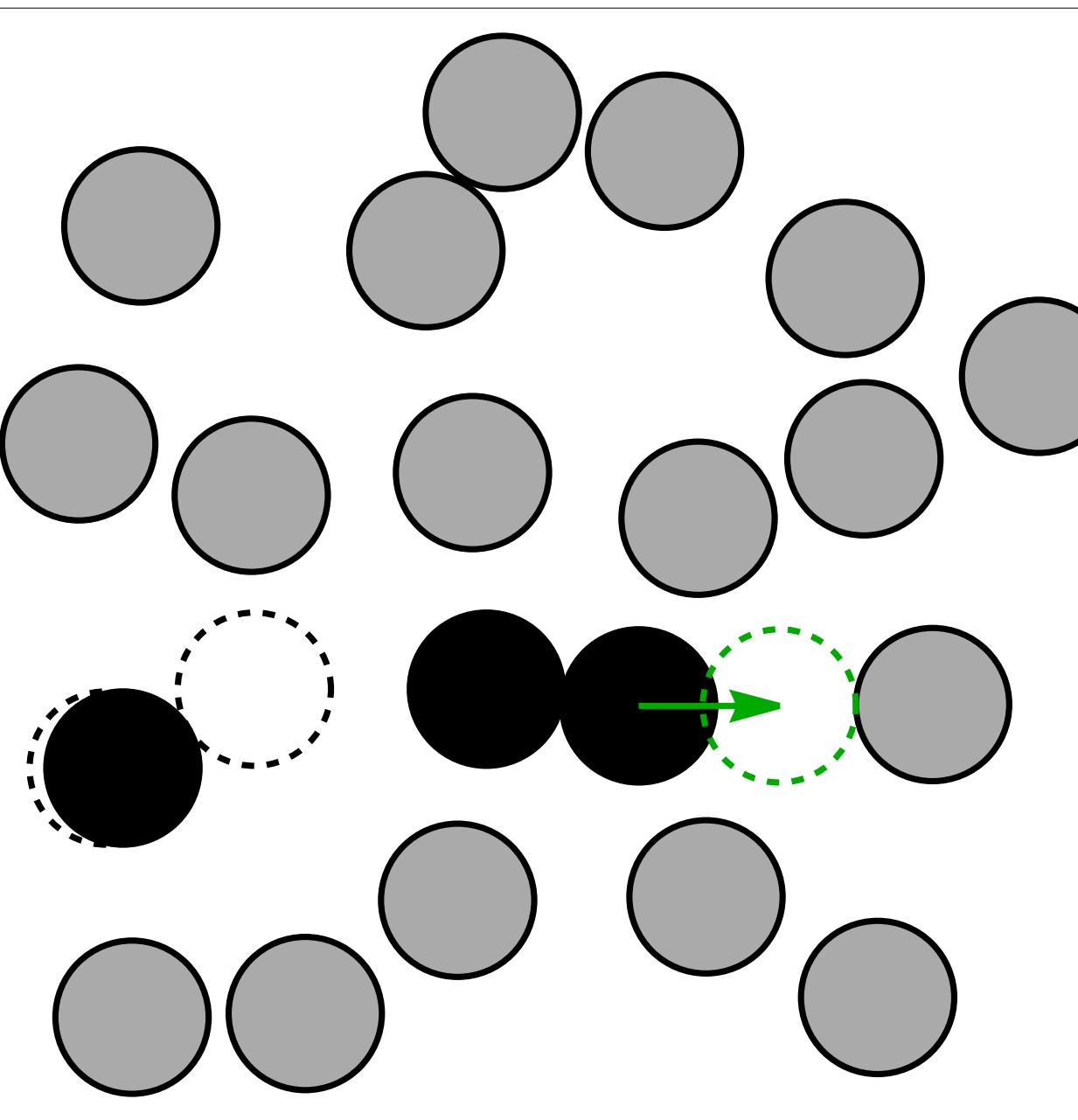
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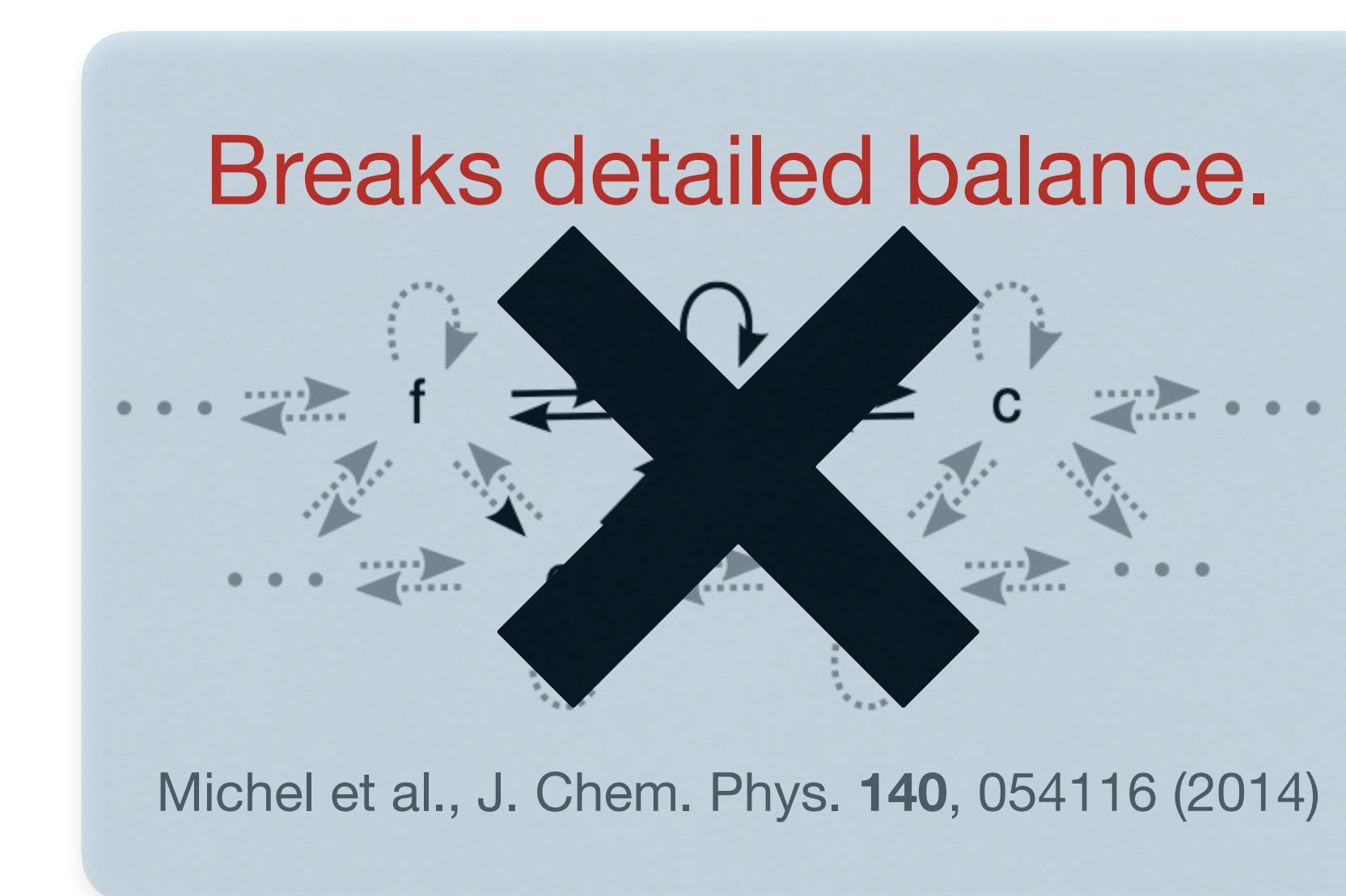
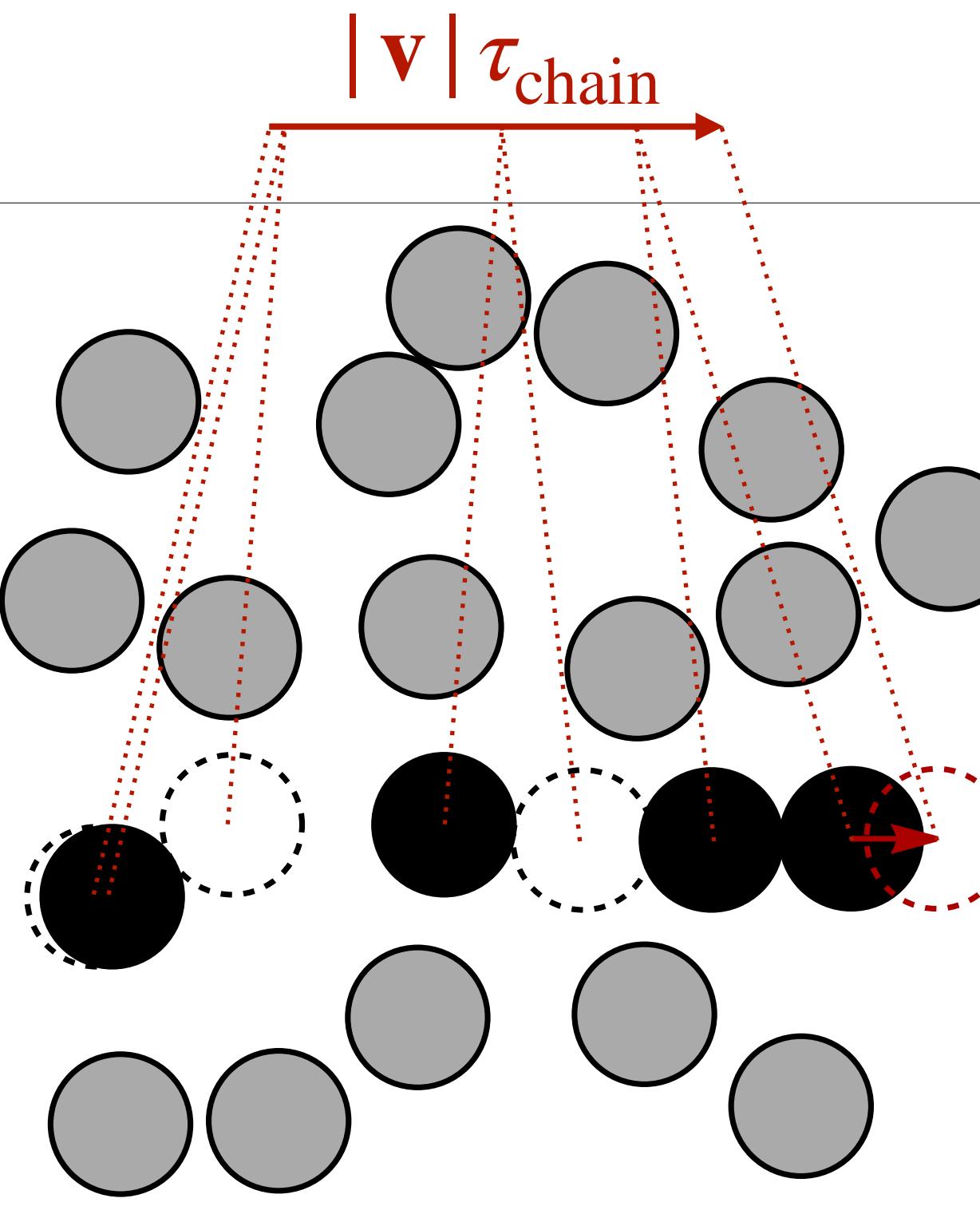
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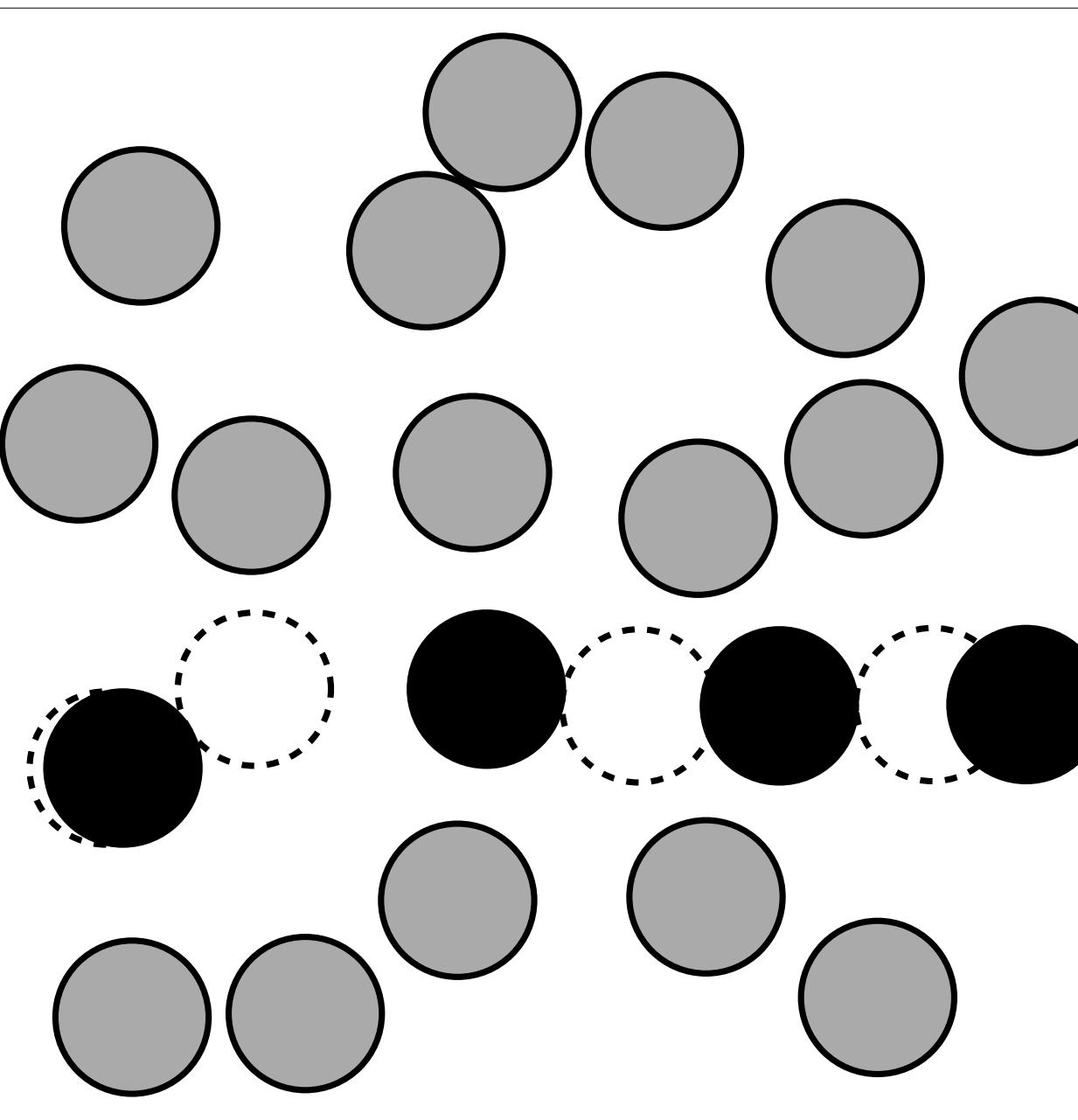
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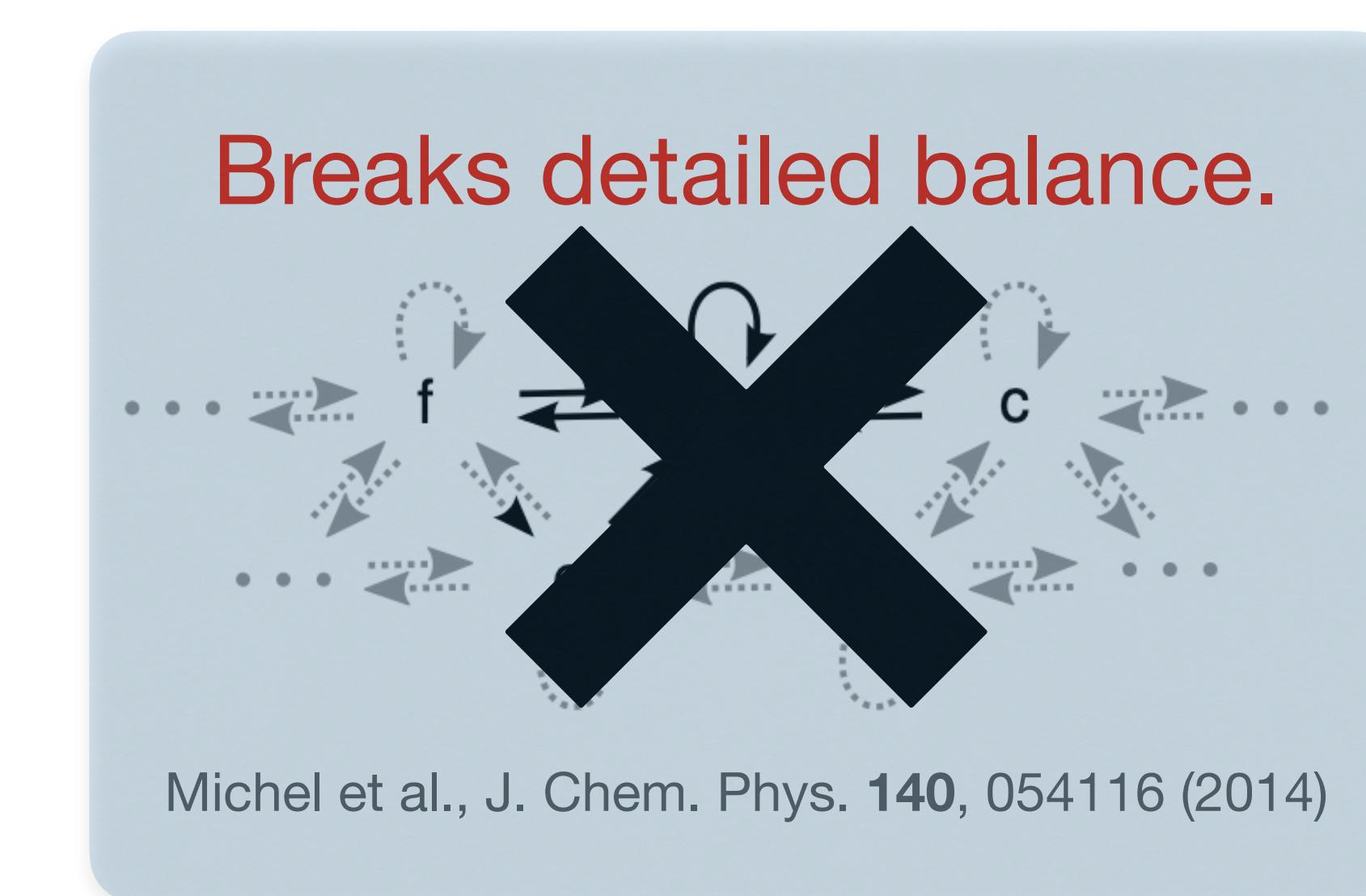
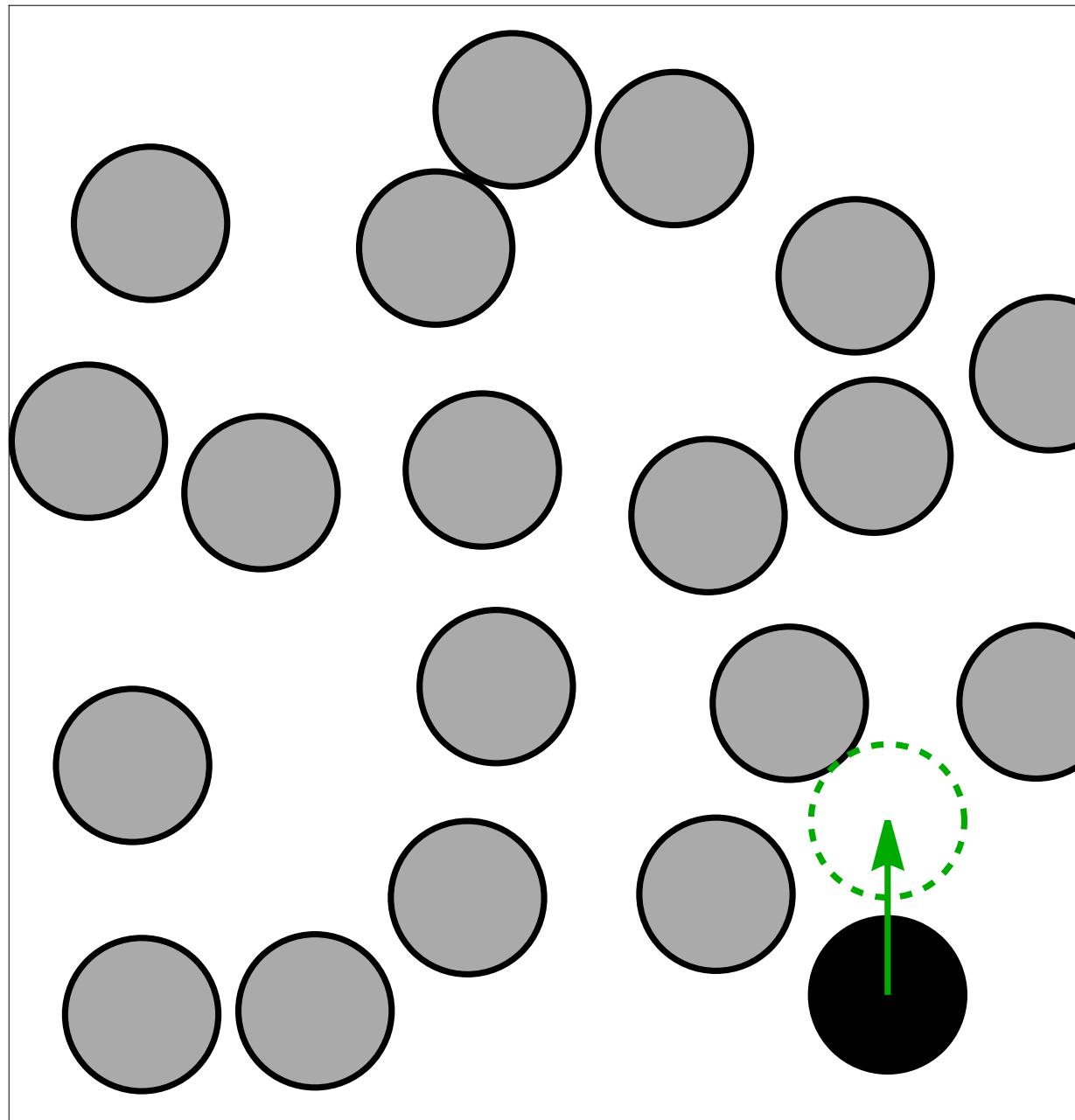
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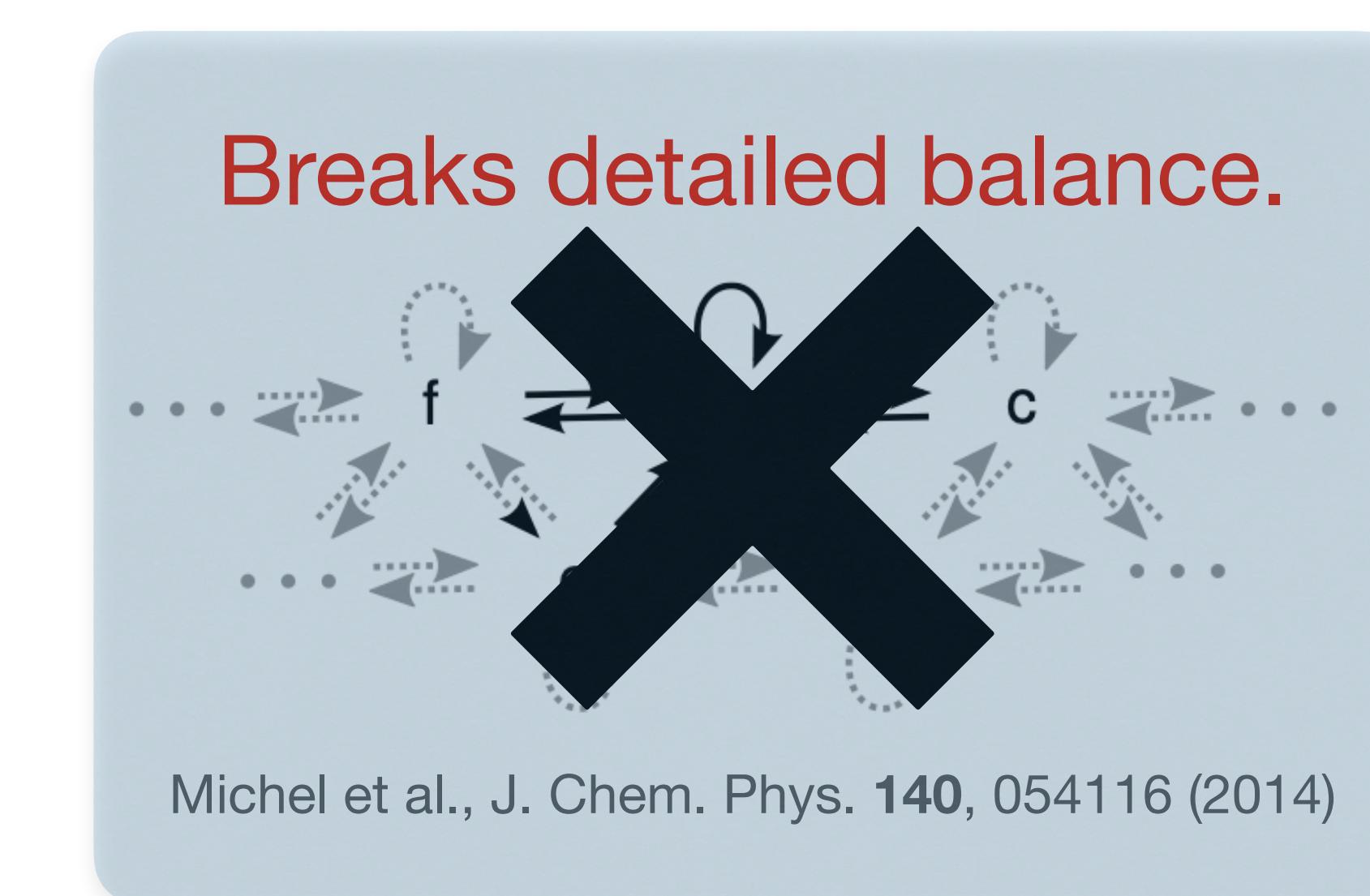
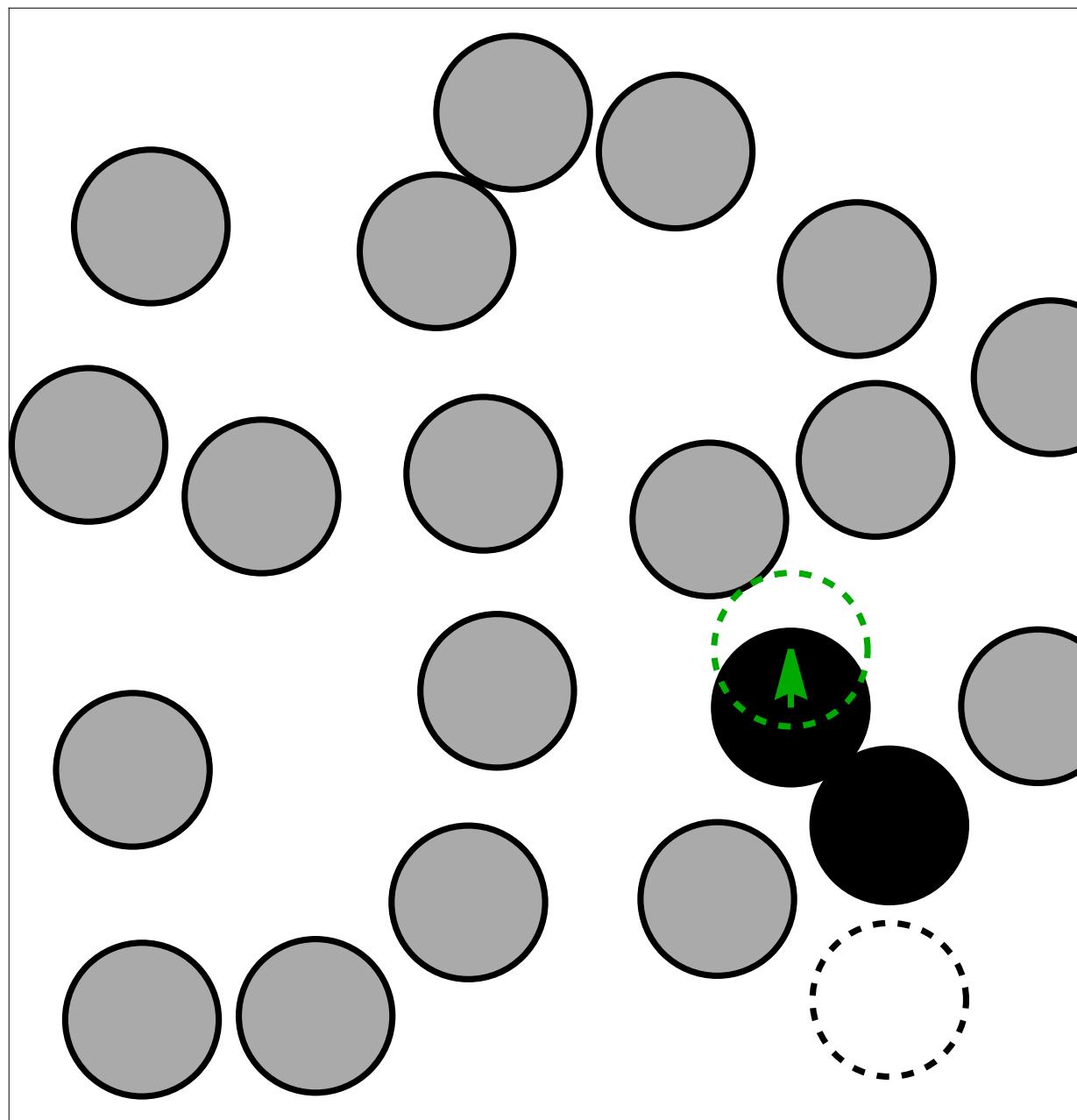
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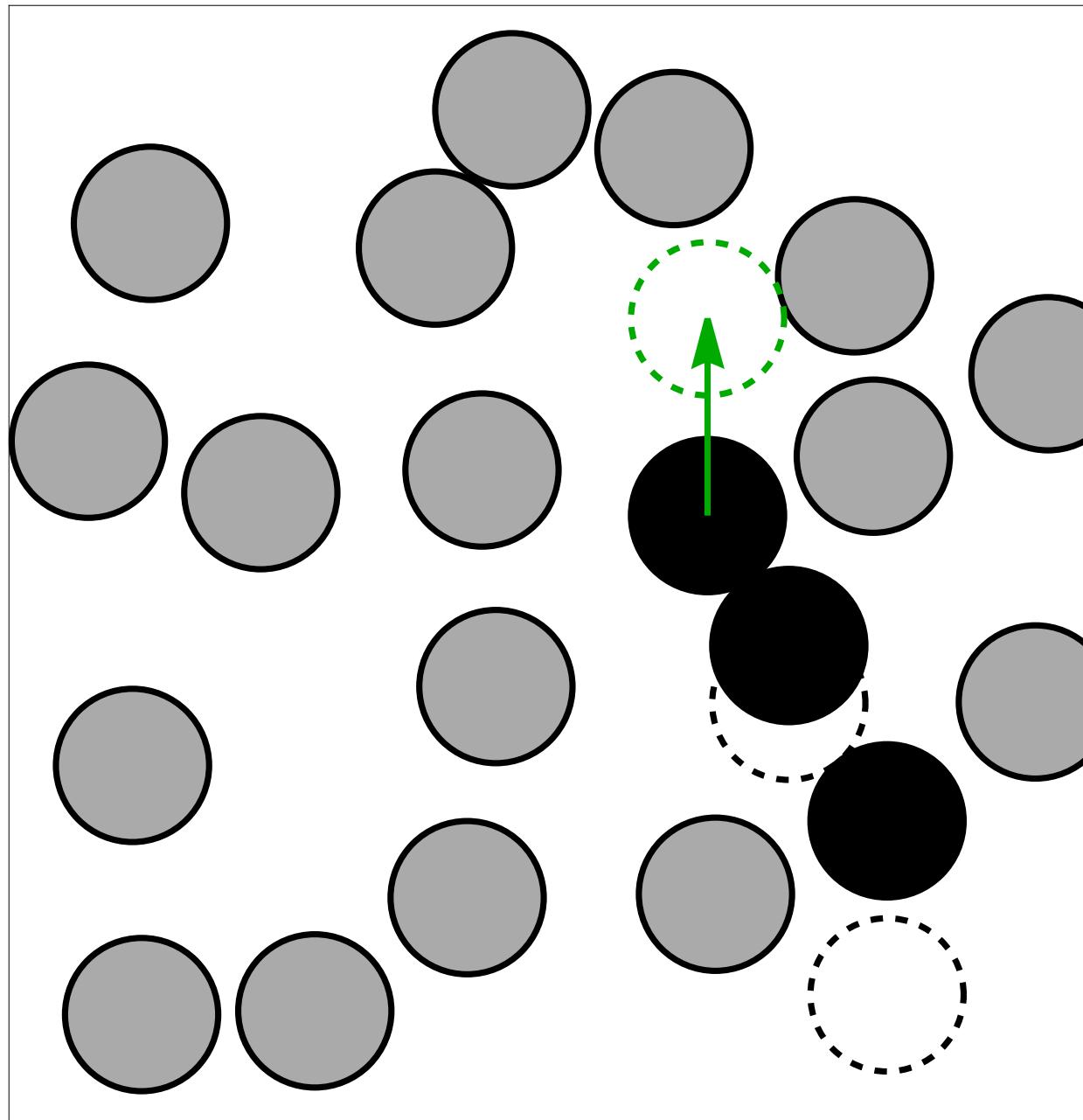
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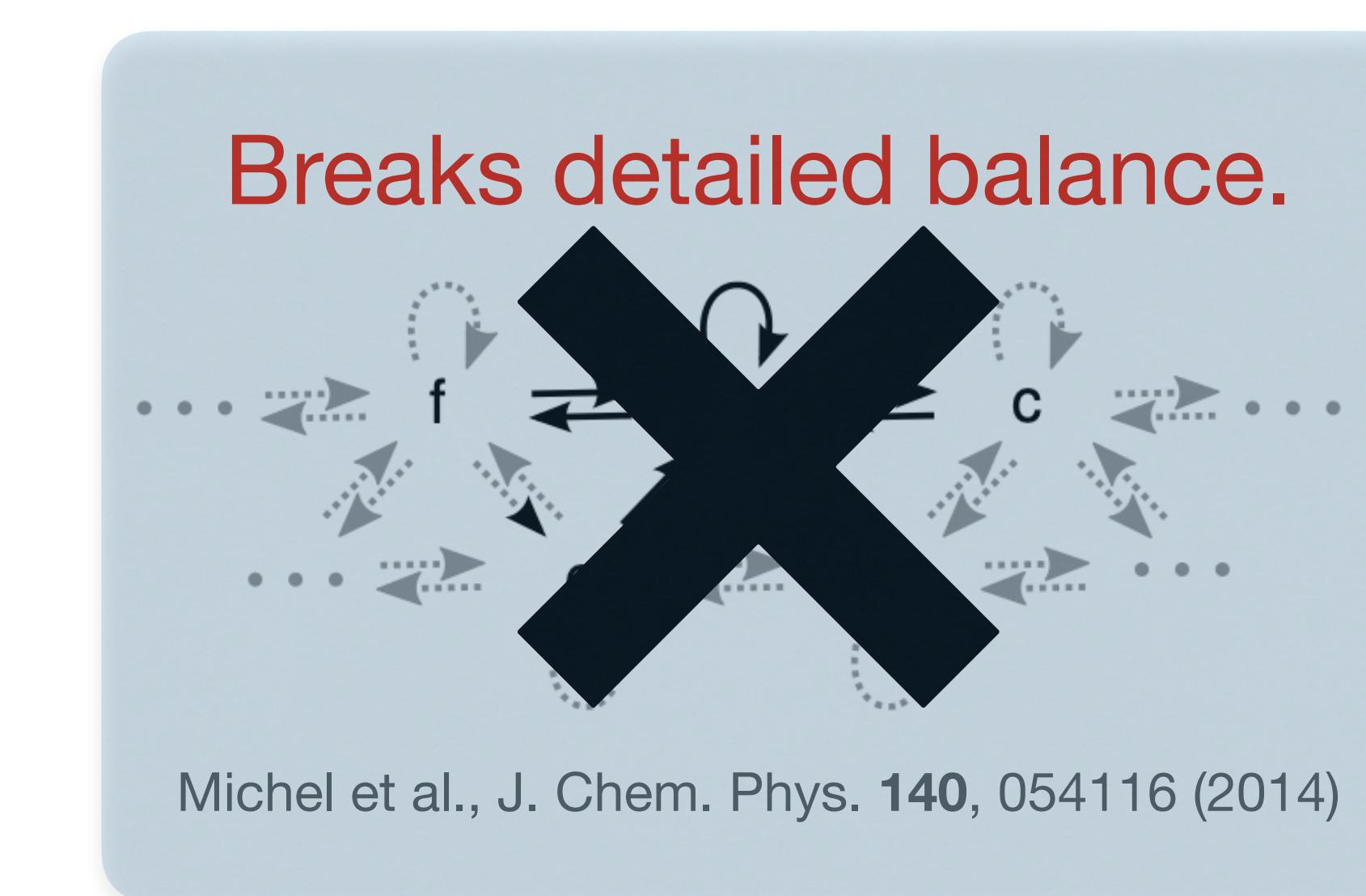
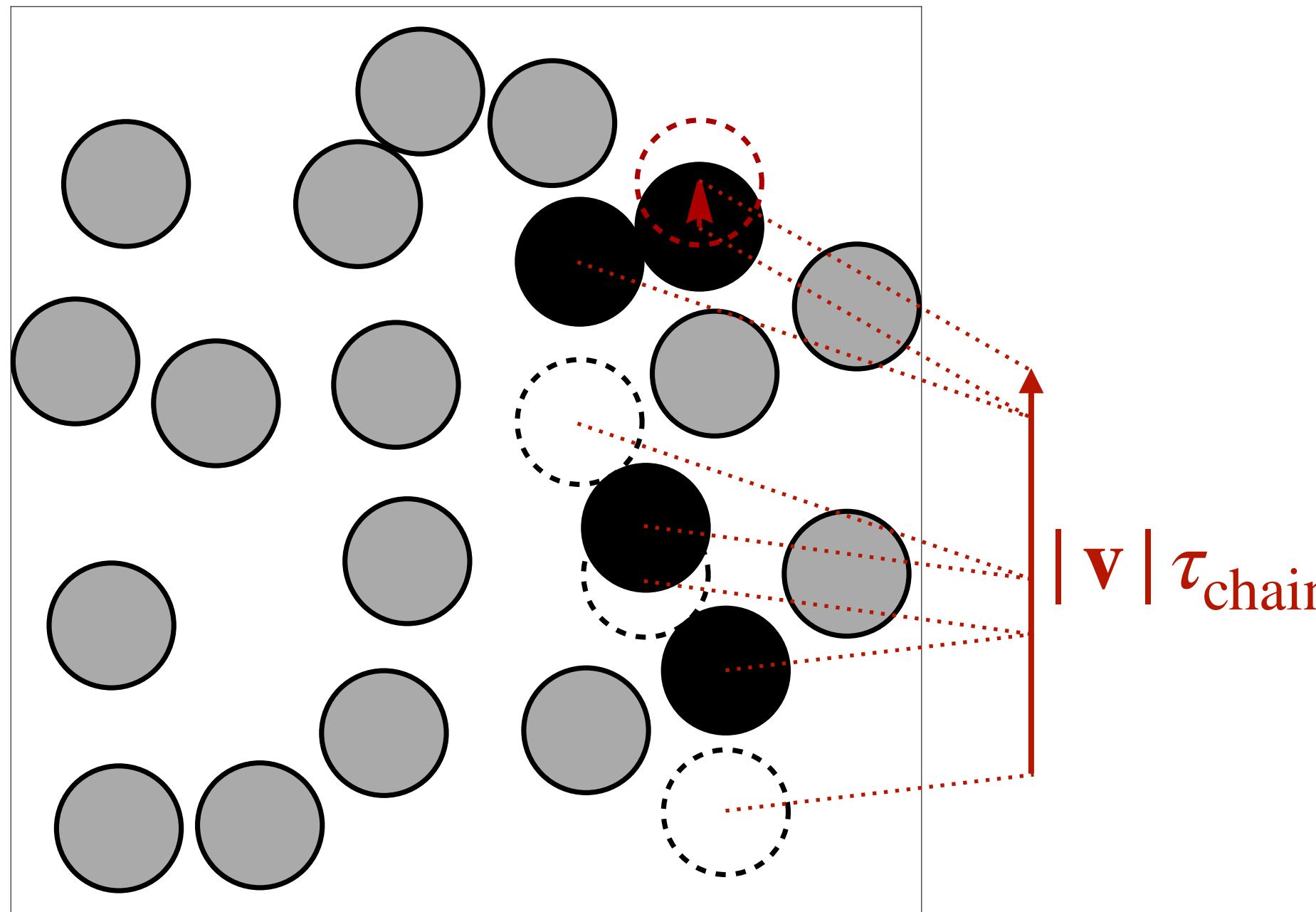
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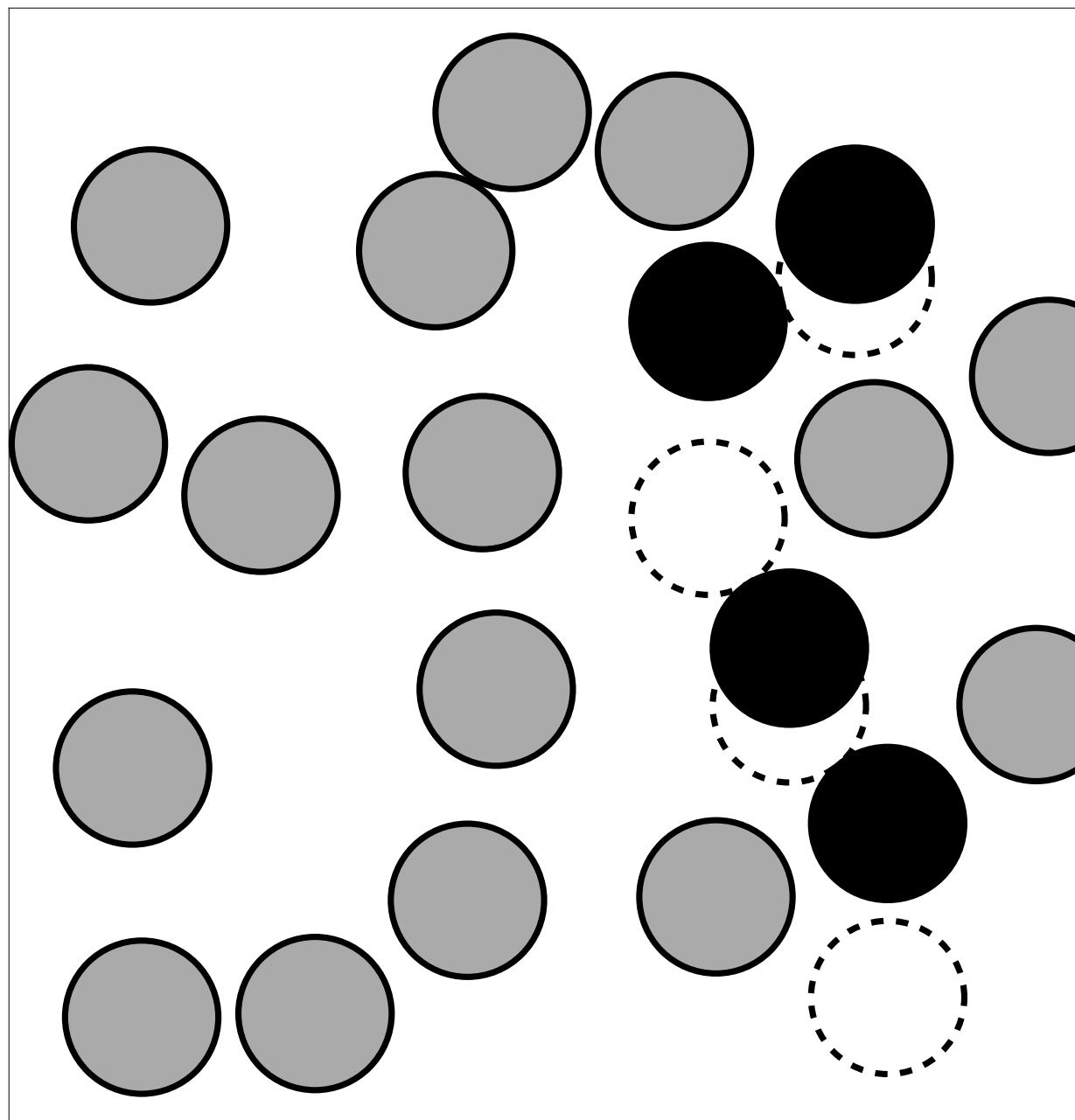
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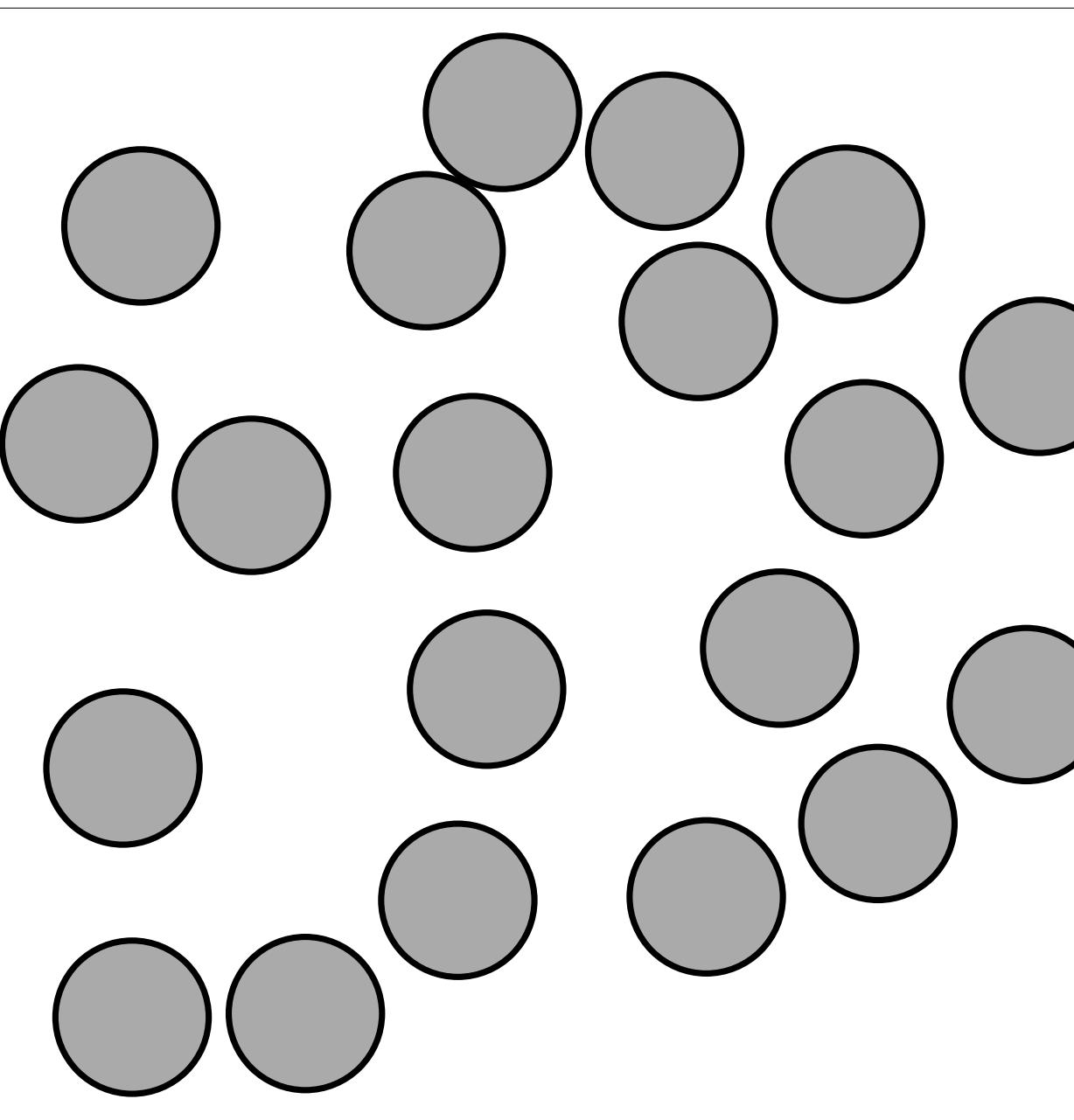
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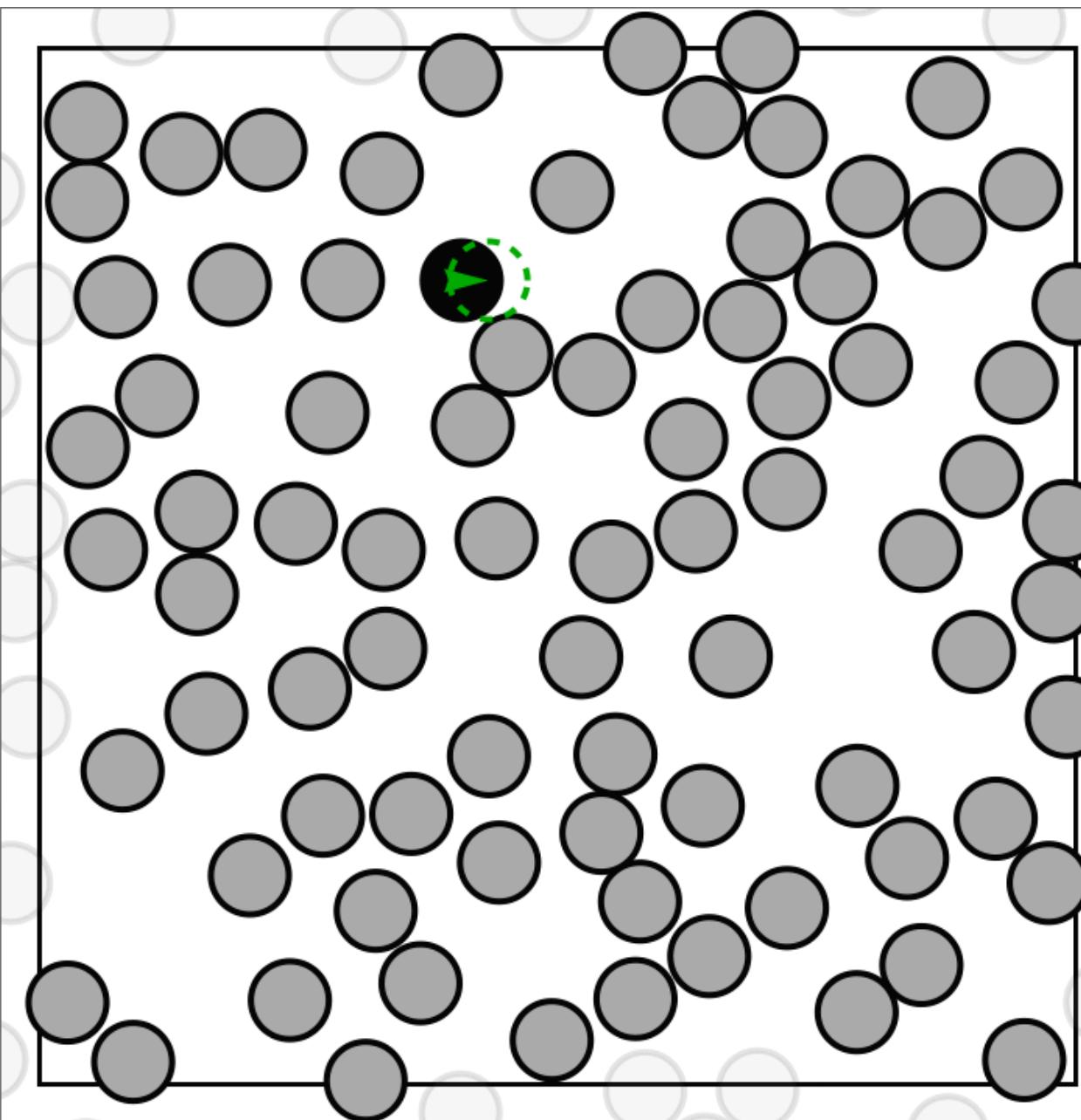
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5. Change velocity to  $\mathbf{v} = (0, 1)^T$  and repeat.



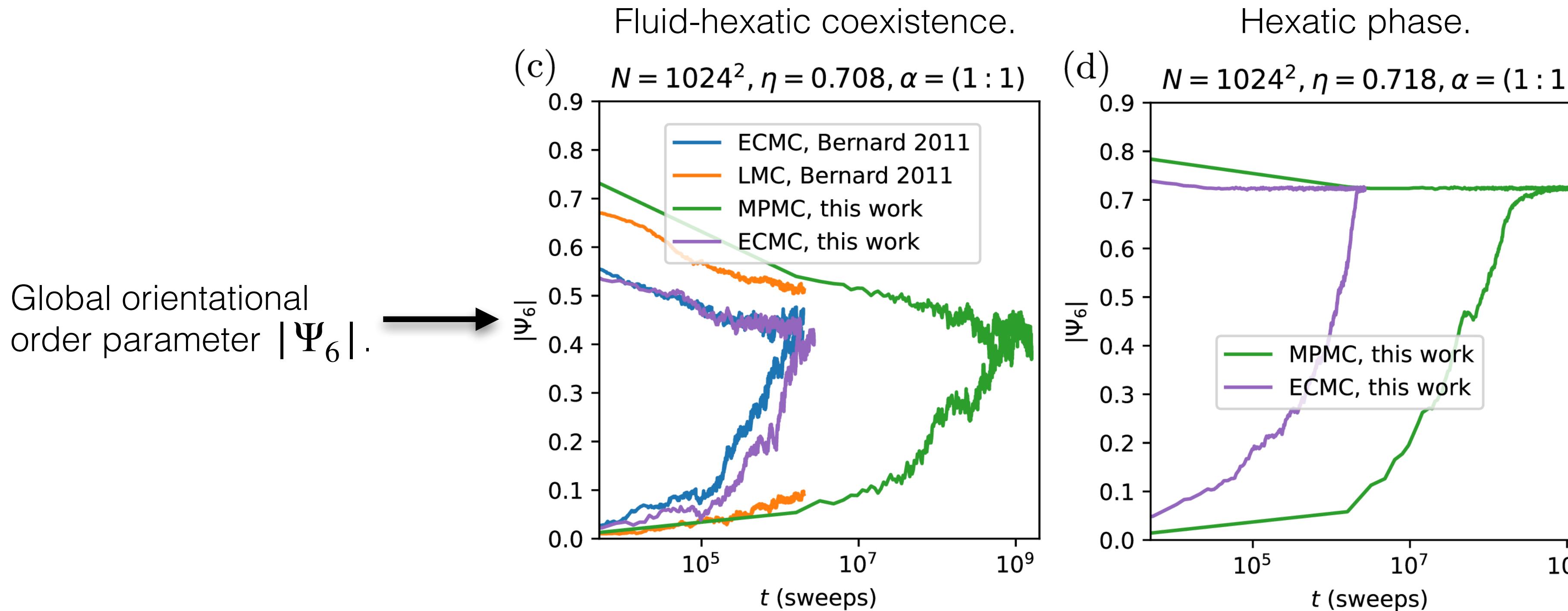
Breaks detailed balance.



Michel et al., J. Chem. Phys. 140, 054116 (2014)

# Phase Transition in the Hard-Disk Model

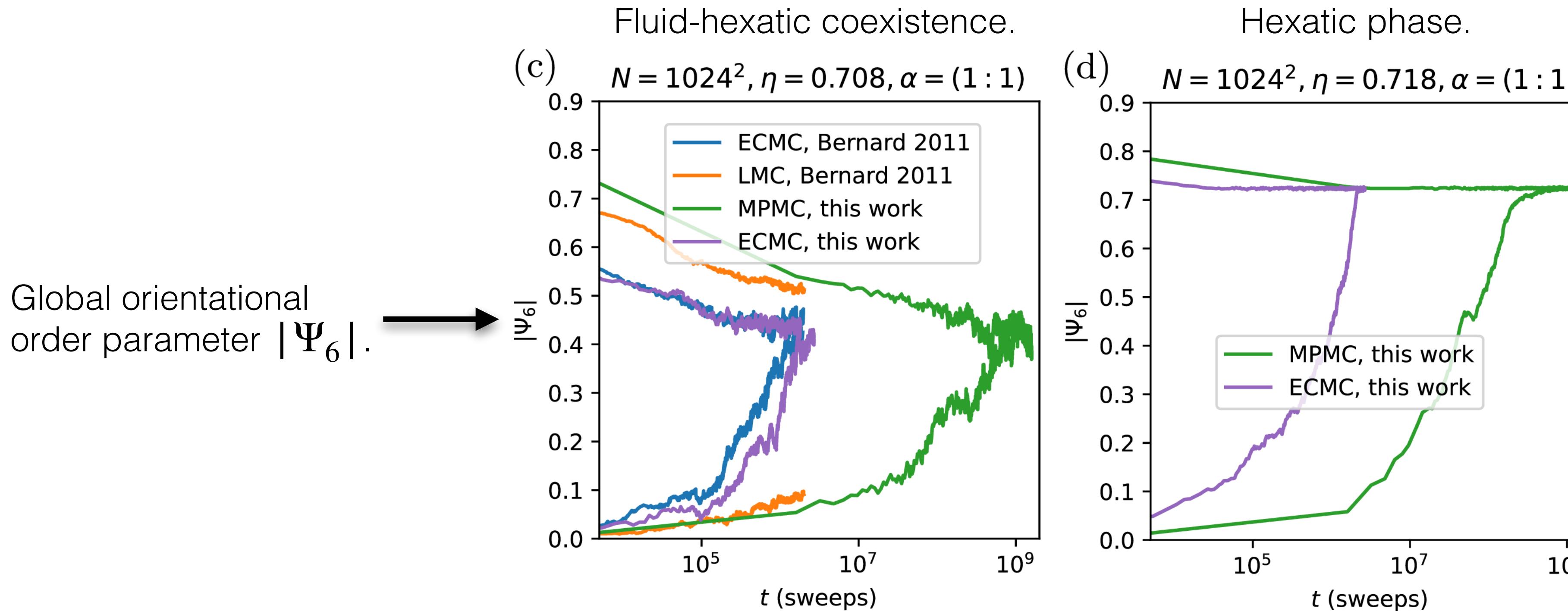
- For hard-disk model at high densities: Non-reversible straight ECMC about three orders of magnitude faster than reversible Metropolis algorithm (LMC).



Li, Nishikawa, PH, Carillo, Maggs, and Krauth, arXiv:2207.07715 (2022, manuscript submitted for publication)

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- Required (single-core) CPU time for coalescence:

ECMC:  $\sim 1$  week.

$\Leftrightarrow$

LMC:  $\sim 10$  years.

# Phase Transition in the Hard-Disk Model

PRL 107, 155704 (2011)

PHYSICAL REVIEW LETTERS

week ending  
7 OCTOBER 2011

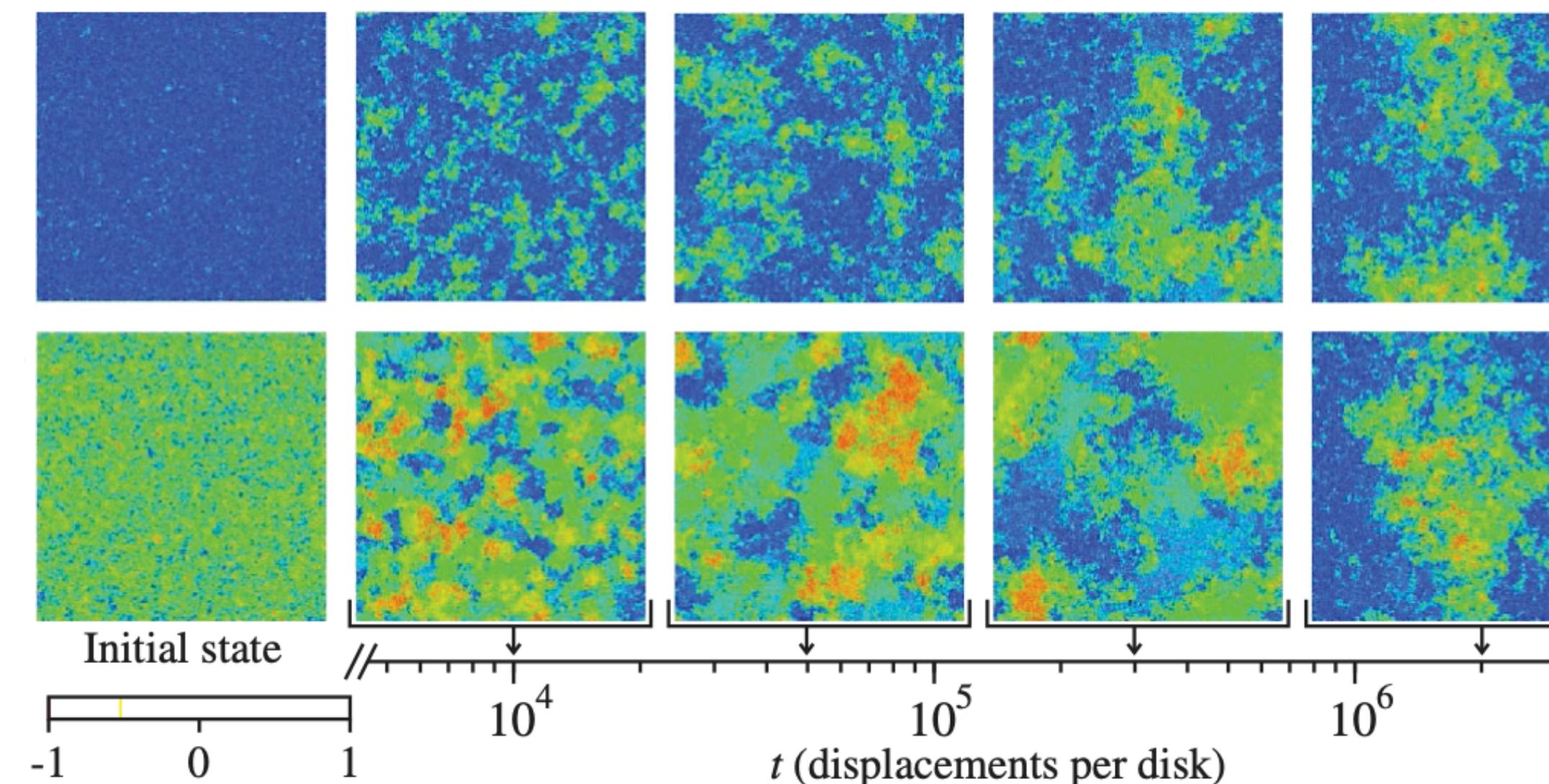
## Two-Step Melting in Two Dimensions: First-Order Liquid-Hexatic Transition

Etienne P. Bernard\* and Werner Krauth†

*Laboratoire de Physique Statistique Ecole Normale Supérieure, UPMC, CNRS 24 rue Lhomond, 75231 Paris Cedex 05, France*

(Received 6 July 2011; published 7 October 2011)

Melting in two spatial dimensions, as realized in thin films or at interfaces, represents one of the most fascinating phase transitions in nature, but it remains poorly understood. Even for the fundamental hard-disk model, the melting mechanism has not been agreed upon after 50 years of studies. A recent Monte Carlo algorithm allows us to thermalize systems large enough to access the thermodynamic regime. We show that melting in hard disks proceeds in two steps with a liquid phase, a hexatic phase, and a solid. The hexatic-solid transition is continuous while, surprisingly, the liquid-hexatic transition is of first order. This melting scenario solves one of the fundamental statistical-physics models, which is at the root of a large body of theoretical, computational, and experimental research.



# Molecular Simulations

## Molecular Dynamics (MD)

- Physical time evolution based on Newton's equations of motion.
- Discretizes time (and possibly space).

## MCMC

- Unphysical time evolution only restricted by global-balance condition.
- Directly samples Boltzmann distribution.

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## MD

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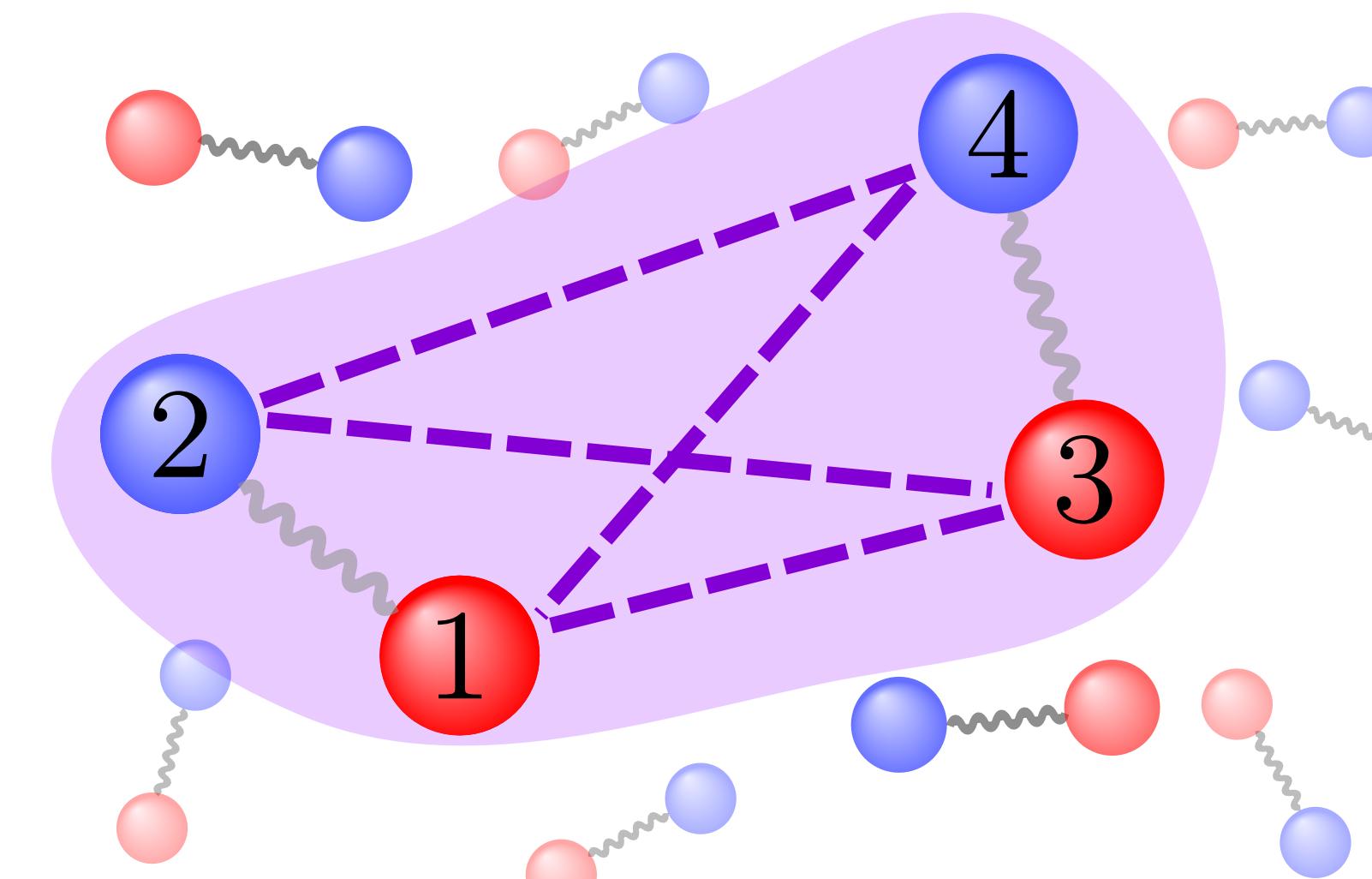
## Reversible MCMC

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**Both disadvantages are overcome by non-reversible ECMC.**

# Generalization of ECMC

- ECMC constructs a **non-reversible rejection-free continuous-time Markov chain**.
- Conceived for systems described by continuous variables.  
→ Here:  $N$  point-like atoms with positions  $\{\mathbf{r}_1, \dots, \mathbf{r}_N\}$ ,  $\mathbf{r}_i \in \mathbb{R}^d$ .
- Exploits translational symmetry of interaction potentials in molecular systems.



Faulkner et al., J. Chem. Phys. 149, 064113 (2018)

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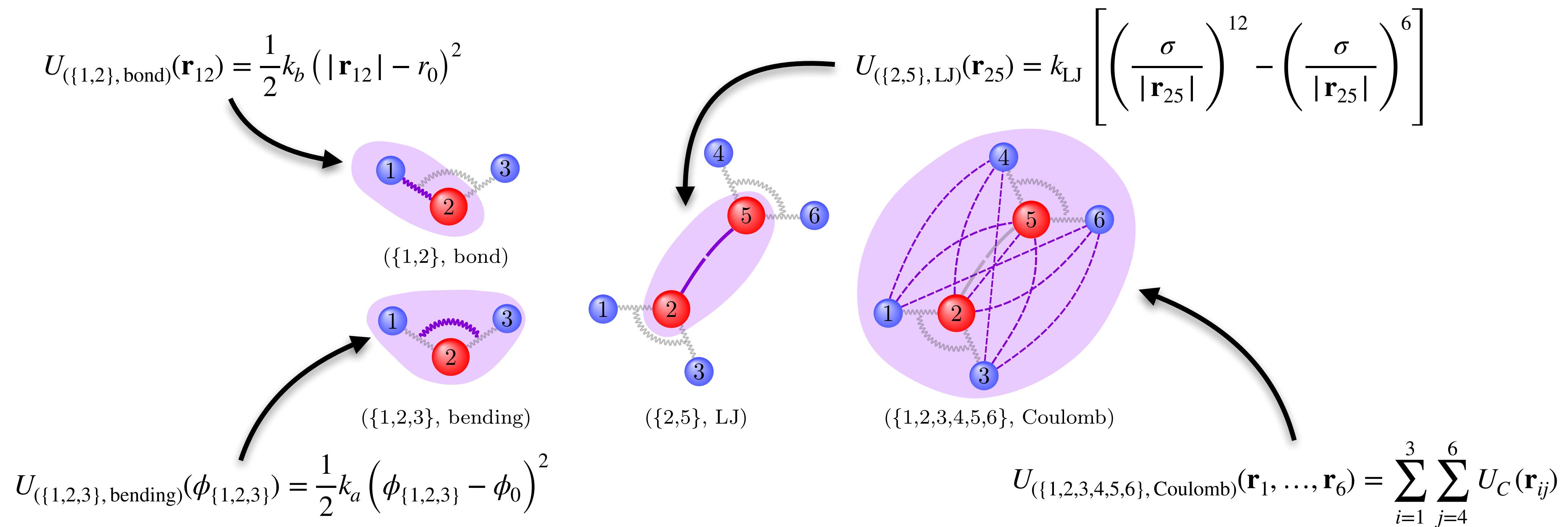
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- Exploits translational symmetry of interaction potentials in molecular systems.
- ECMC relies on three concepts:
  1. **Factorized Metropolis filter** accepts/rejects proposed configurations.
  2. **Lifting framework** proposes new configurations and solves rejections.
  3. **Event-driven implementation**.

# Factorized Metropolis Filter

- Decompose total potential  $U$  into sum over factor potentials  $U_M$  of factors  $M$ :

$$U(c = \{\mathbf{r}_1, \dots, \mathbf{r}_N\}) = \sum_M U_M(c_M = \{\mathbf{r}_i : i \in M\}) \longrightarrow \pi_B(c) = \frac{1}{Z} \prod_M e^{-\beta U_M(c_M)}$$

- SPC/Fw water model:



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- Replace original Metropolis filter that accepts a change of configuration  $c \rightarrow c'$ :

$$A^{\text{Met}}(c \rightarrow c') = \min \left[ 1, \prod_M e^{-\beta \Delta U_M(c_M \rightarrow c'_M)} \right] \longrightarrow A^{\text{Fact}}(c \rightarrow c') = \prod_M \min \left[ 1, e^{-\beta \Delta U_M(c_M \rightarrow c'_M)} \right]$$

Original Metropolis filter.

Factorized Metropolis filter.

- Formulates **consensus principle**, new configuration is accepted from all factors independently:

$$X^{\text{Fact}}(c \rightarrow c') = \bigwedge_M X_M(c_M \rightarrow c'_M)$$

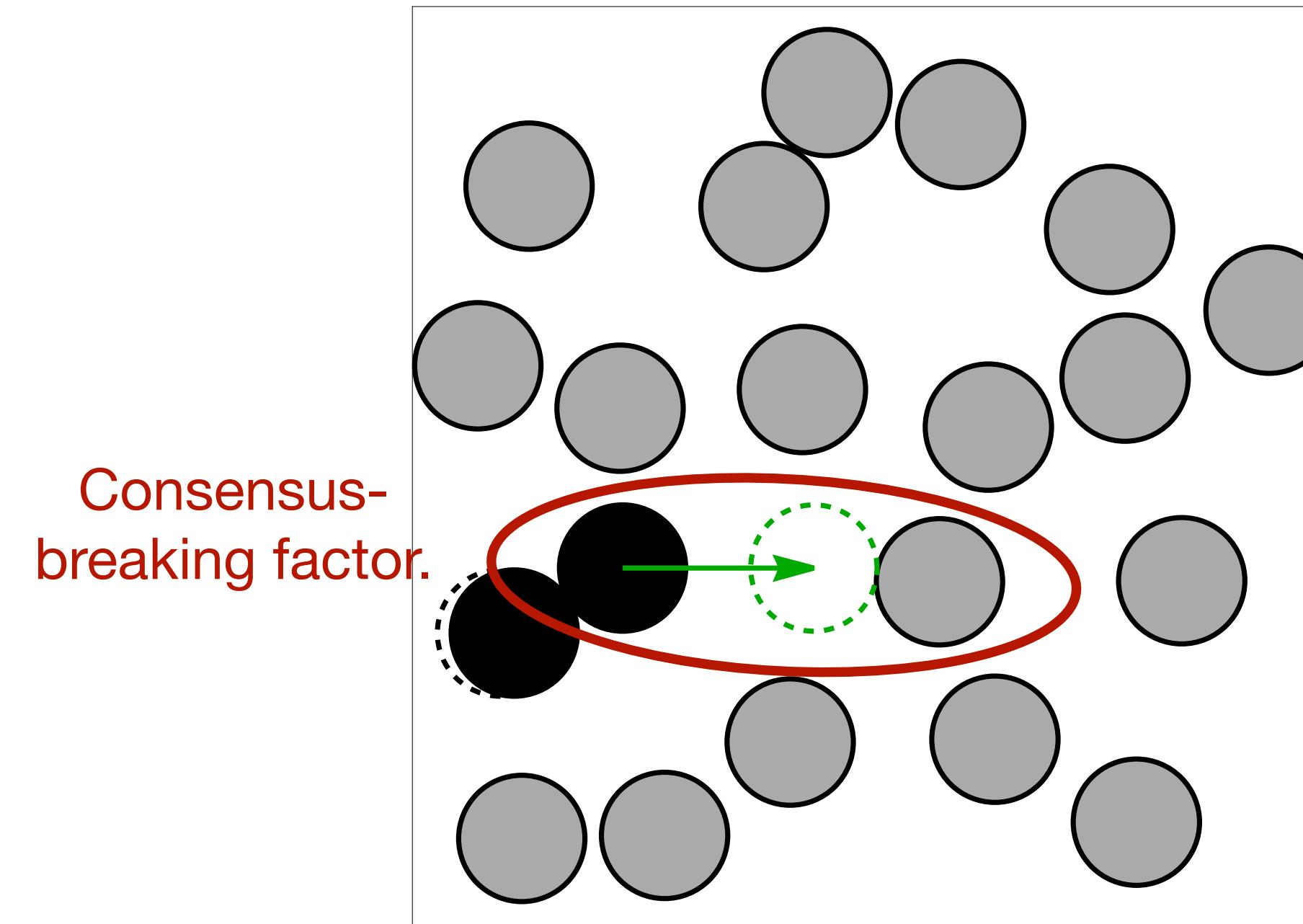
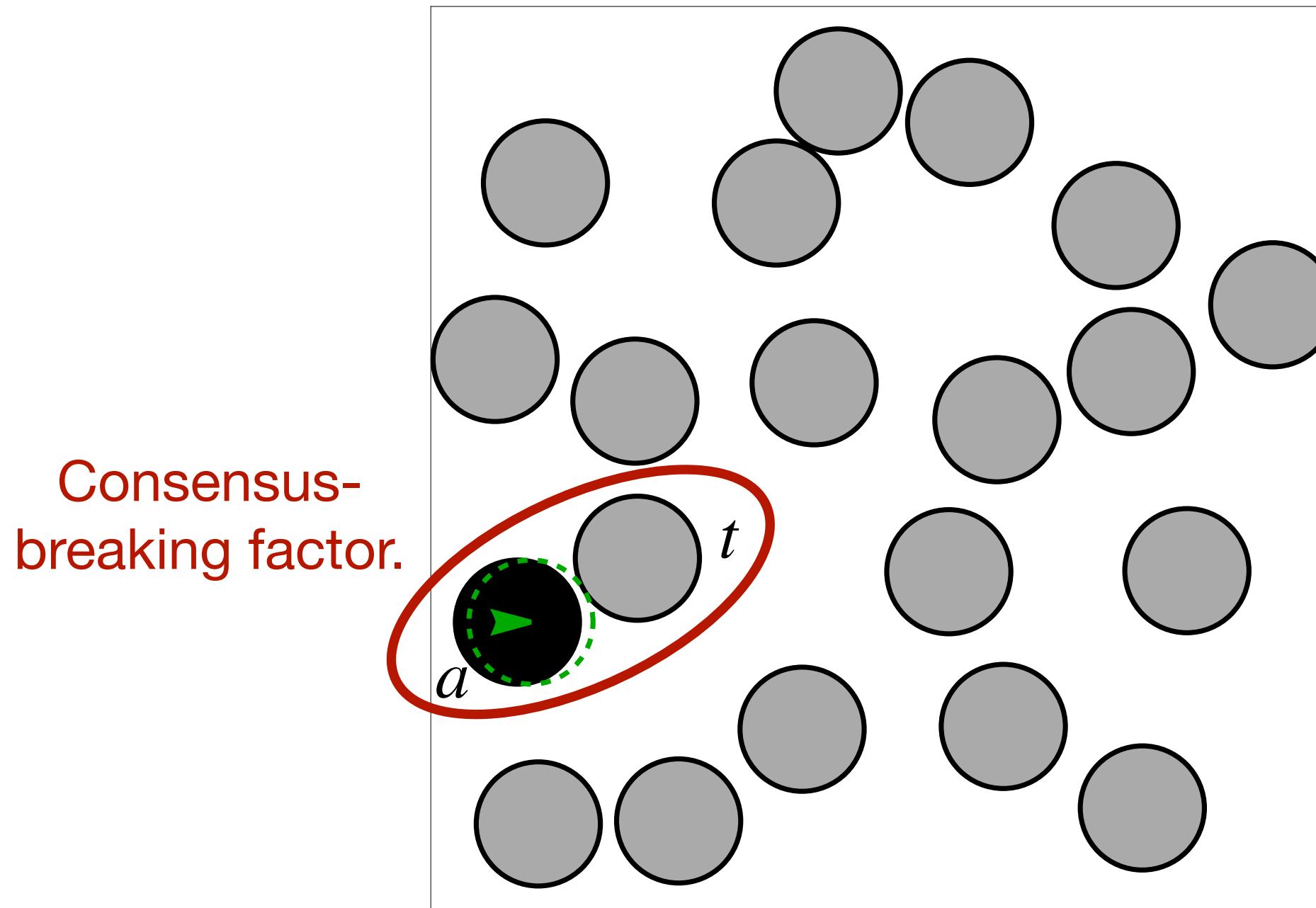
Whether the new configuration is accepted by all factors

$$X_M(c_M \rightarrow c'_M) = \begin{cases} \text{True} & \text{if } \text{ran}_M(0,1) < e^{-\beta \Delta U_M(c_M \rightarrow c'_M)} \\ \text{False} & \text{otherwise} \end{cases}$$

Whether the new configuration is accepted by factor  $M$

# Lifting Framework

- Lift configuration  $c \rightarrow (c, \mathbf{v}, a)$  to include active atom  $a$  and its velocity  $\mathbf{v}$ .
- Atom  $a$  moves with  $\mathbf{v}$  until a factor breaks consensus.
- If a translationally symmetric pair potential between  $a$  and  $t$  breaks consensus:  
→ Change active atom from  $a$  to  $t$  in a lifting move.



# Event-Driven Implementation

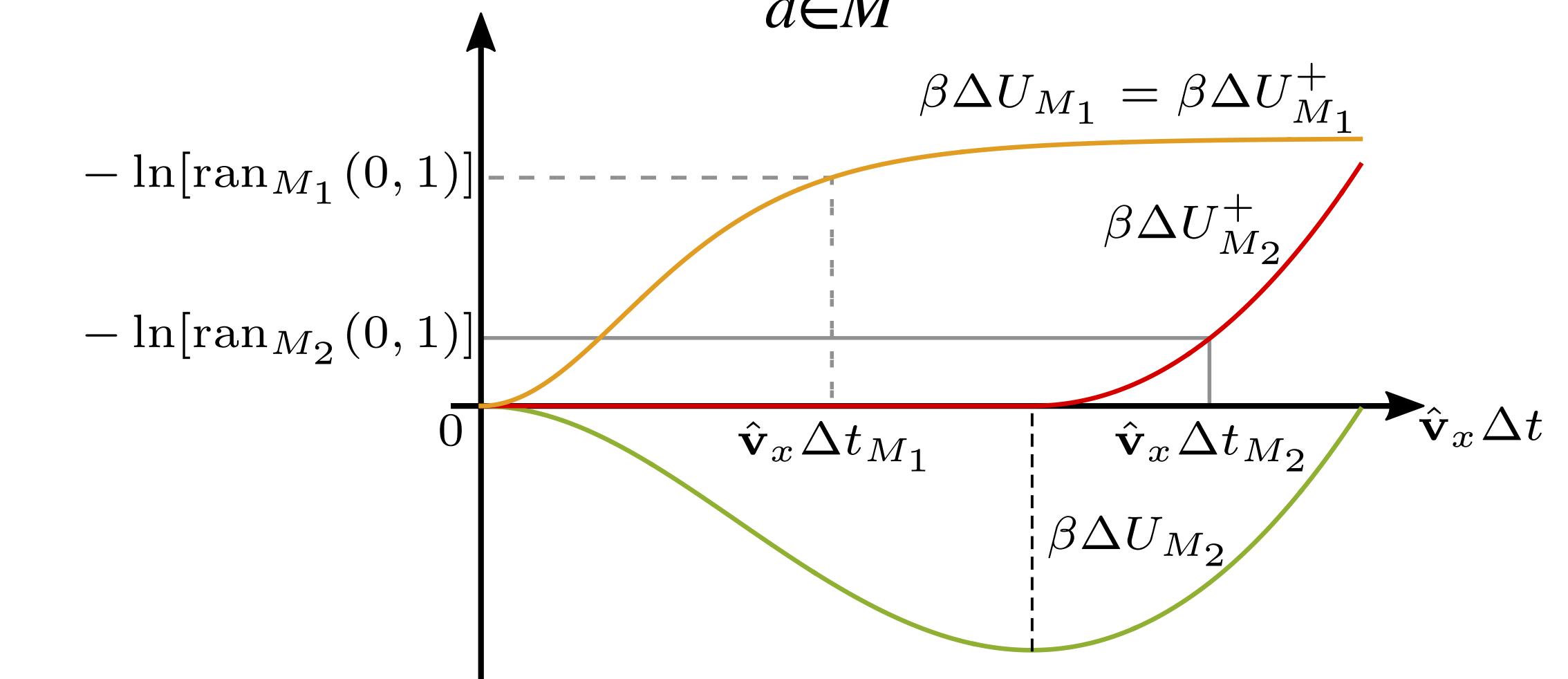
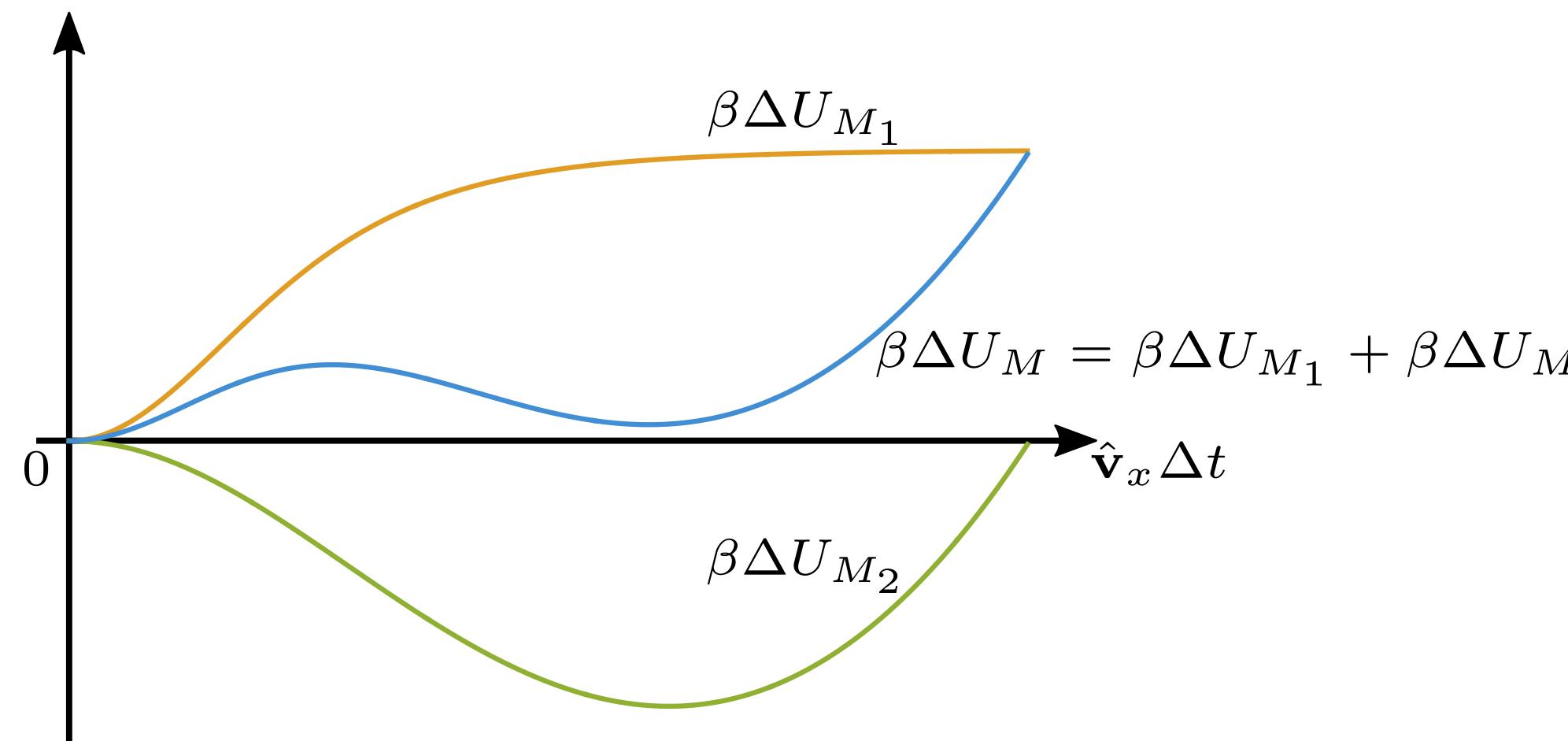
- For every factor  $M$  containing  $a$ , sample a candidate event time  $\Delta t_M$ .

→ Proposed configuration  $c'_M : c_M = \{\mathbf{r}_a, \mathbf{r}_t\} \rightarrow c'_M = \{\mathbf{r}_a + \Delta t_M \mathbf{v}, \mathbf{r}_t\}$ .

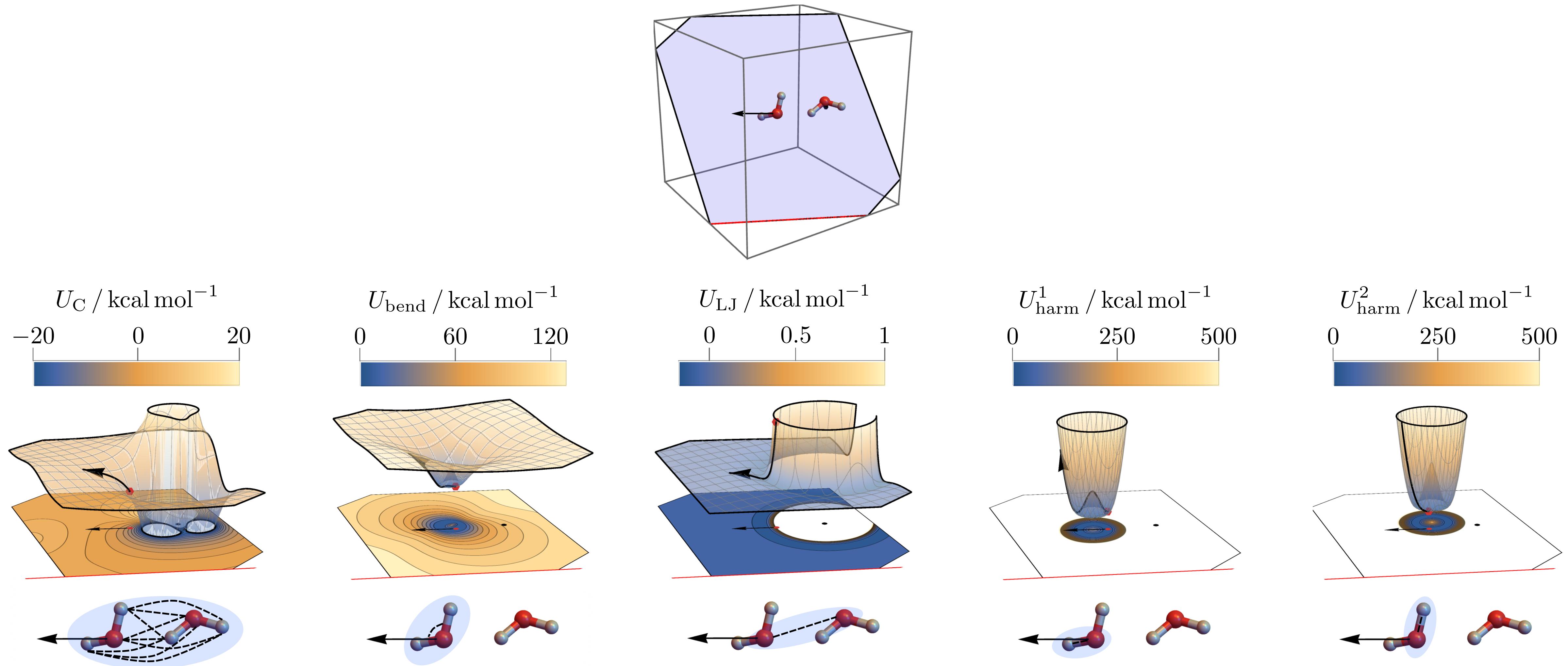
→ Invert equation for  $\Delta t_M$ :  $-\ln [\text{ran}_M(0,1)] = \beta \Delta U_M^+(c_M \rightarrow c'_M)$ .

→ Integrate event rate  $\Delta U_M^+(c_M \rightarrow c'_M) = \int_0^{\Delta t_M} dt \max [0, \mathbf{v} \cdot \nabla_{\mathbf{r}_a} U_M(\mathbf{r}_a + t\mathbf{v}, \mathbf{r}_t)]$ .

- Event takes place at minimum candidate event time  $\Delta t_{\text{Event}} = \min_{a \in M} \Delta t_M$ .

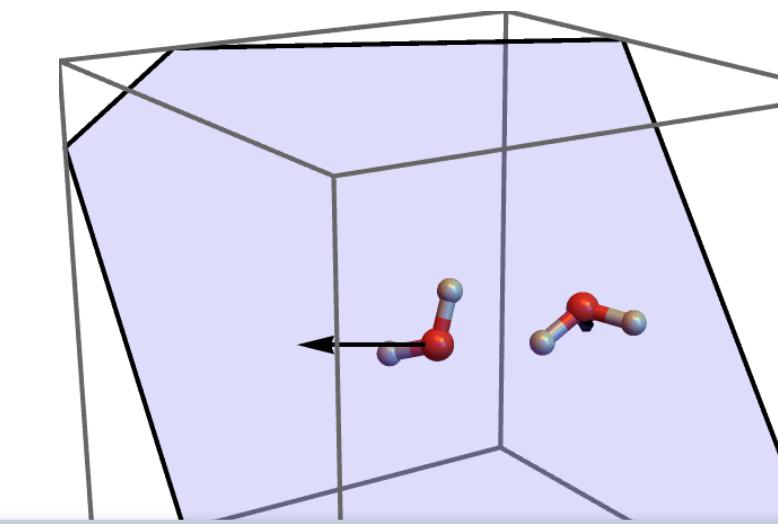


# ECMC for SPC/Fw Water Model



PH, Maggs, and Krauth, *Bringing the Power of Monte Carlo methods to Long-Range-Interacting Molecular Systems* (2022, manuscript in preparation)

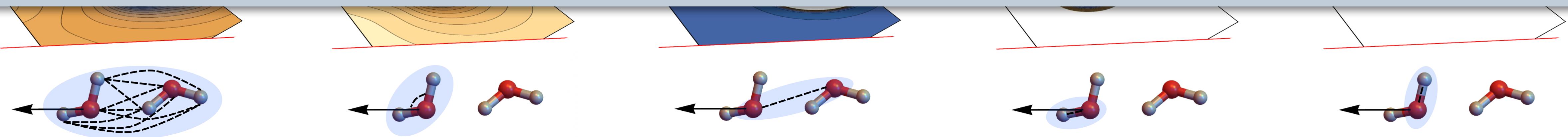
# ECMC for SPC/Fw Water Model



## Problems

- Integration and inversion of event rate can be tedious.
- In long-range-interacting  $N$ -body systems, consensus considers  $\mathcal{O}(N)$  factors per event.

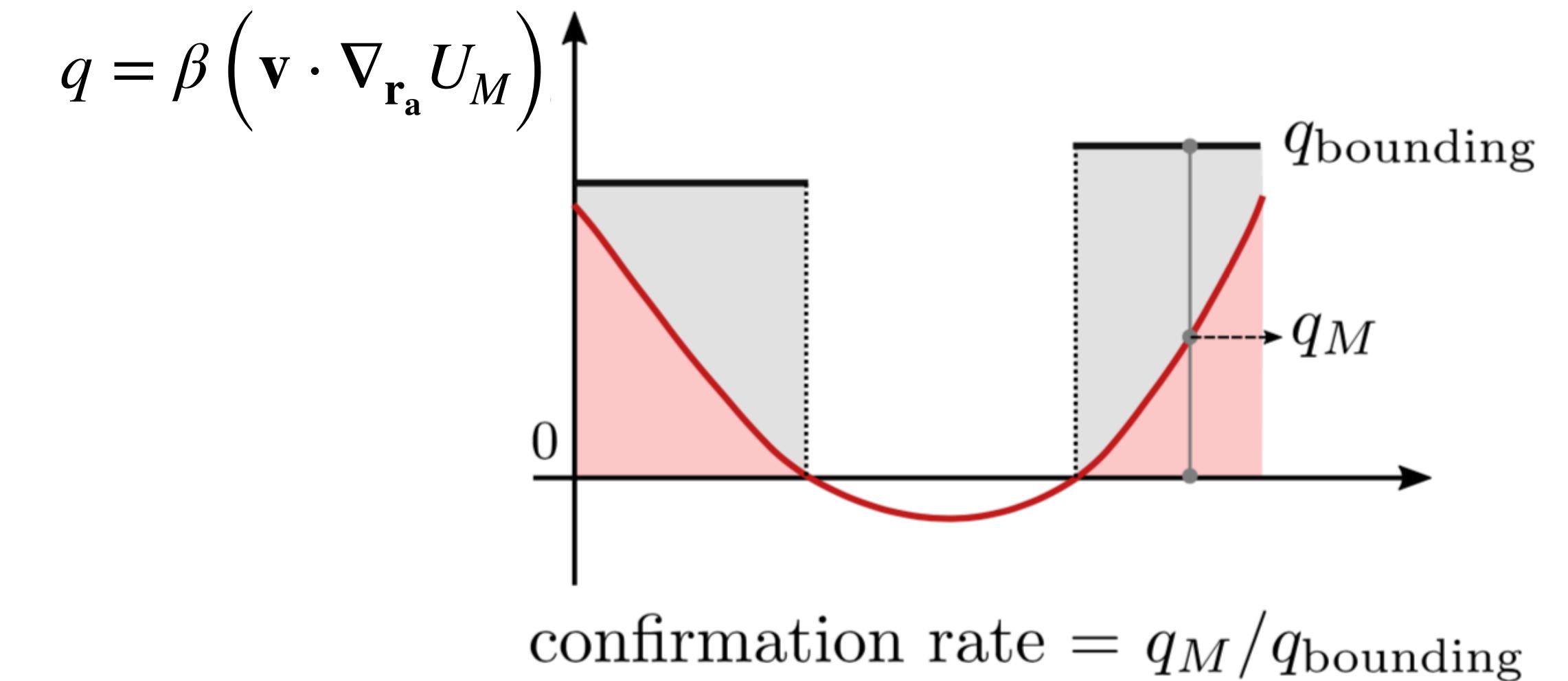
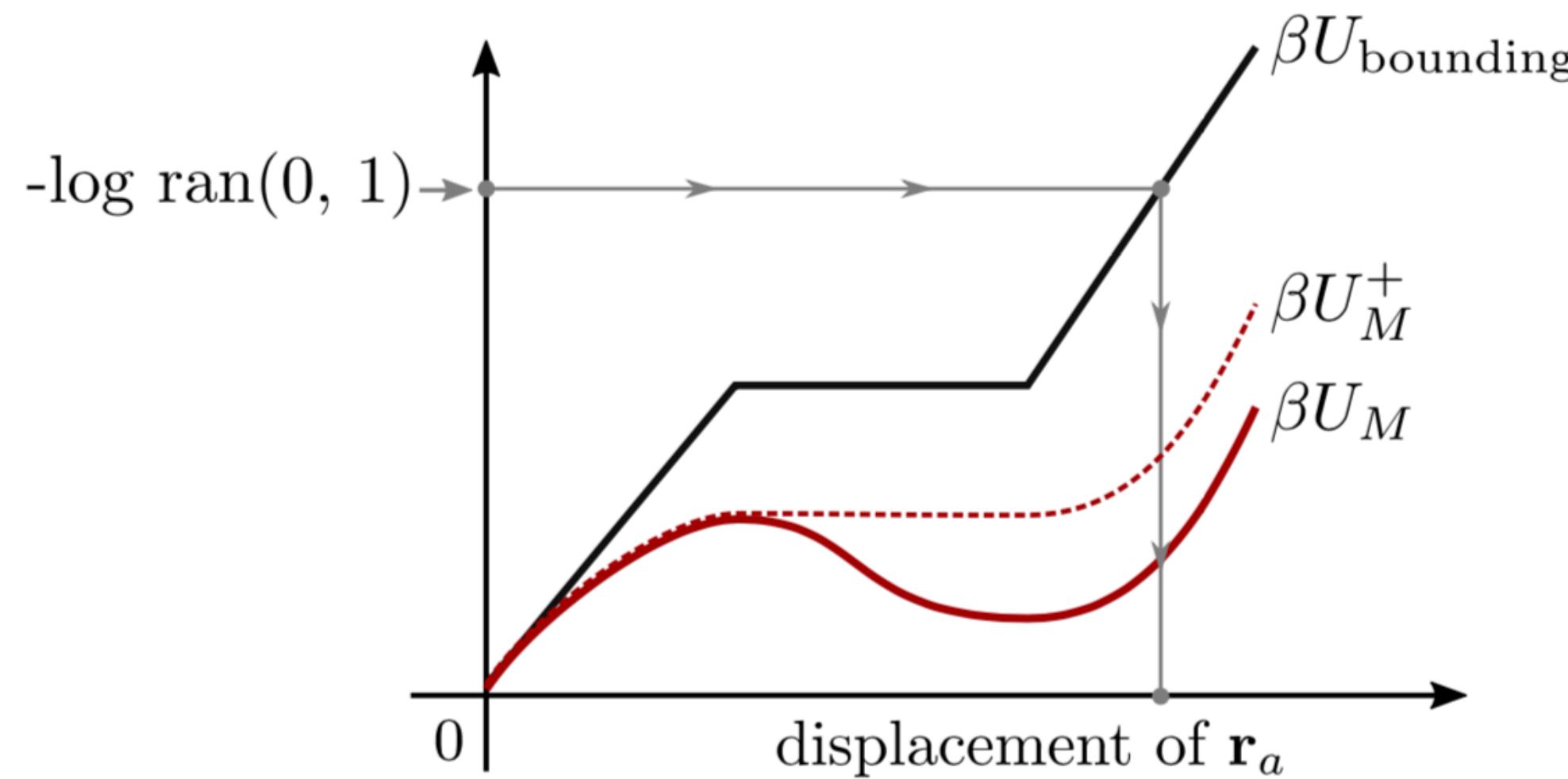
→ Problematic long-range merged-image Coulomb potential.



PH, Maggs, and Krauth, *Bringing the Power of Monte Carlo methods to Long-Range-Interacting Molecular Systems* (2022, manuscript in preparation)

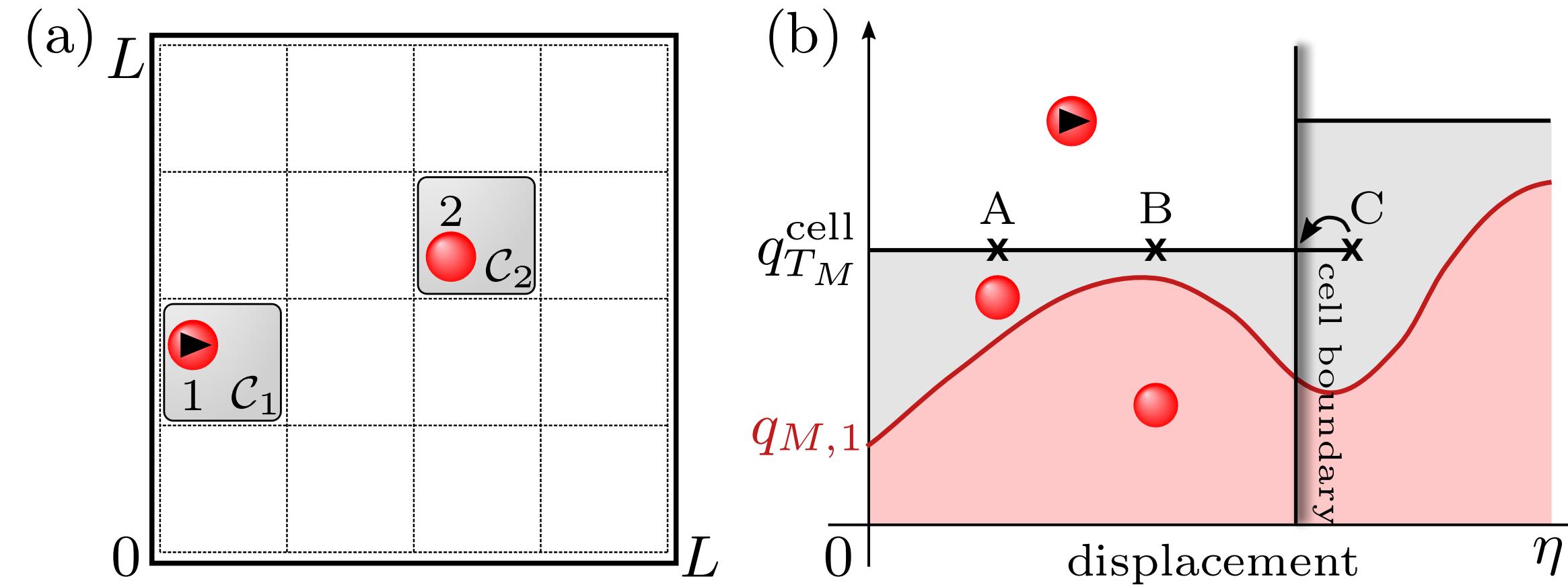
# Bounding Potentials

- Bounding potentials bound the event rate of their factor potential.  
→ Events of bounding potentials have to be confirmed.



# Cell-Veto Algorithm

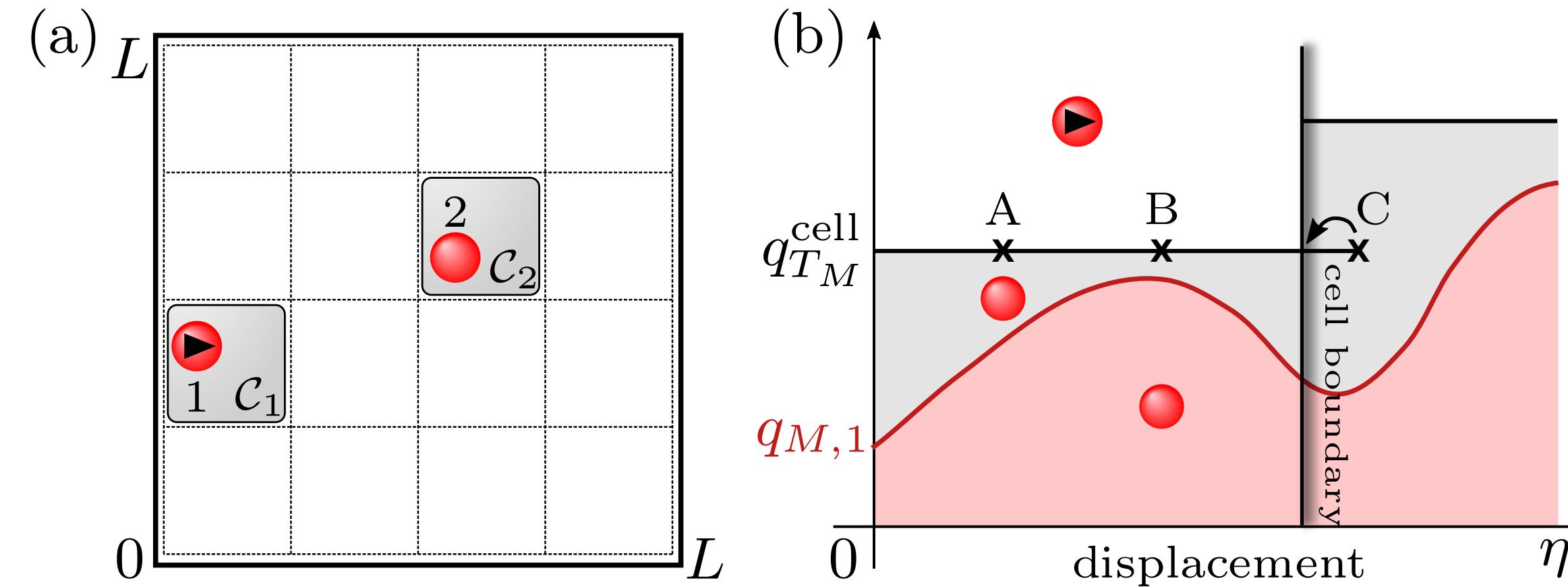
- Cell-based bounding potentials can be precomputed.



Faulkner et al., J. Chem. Phys. 149, 064113 (2018)

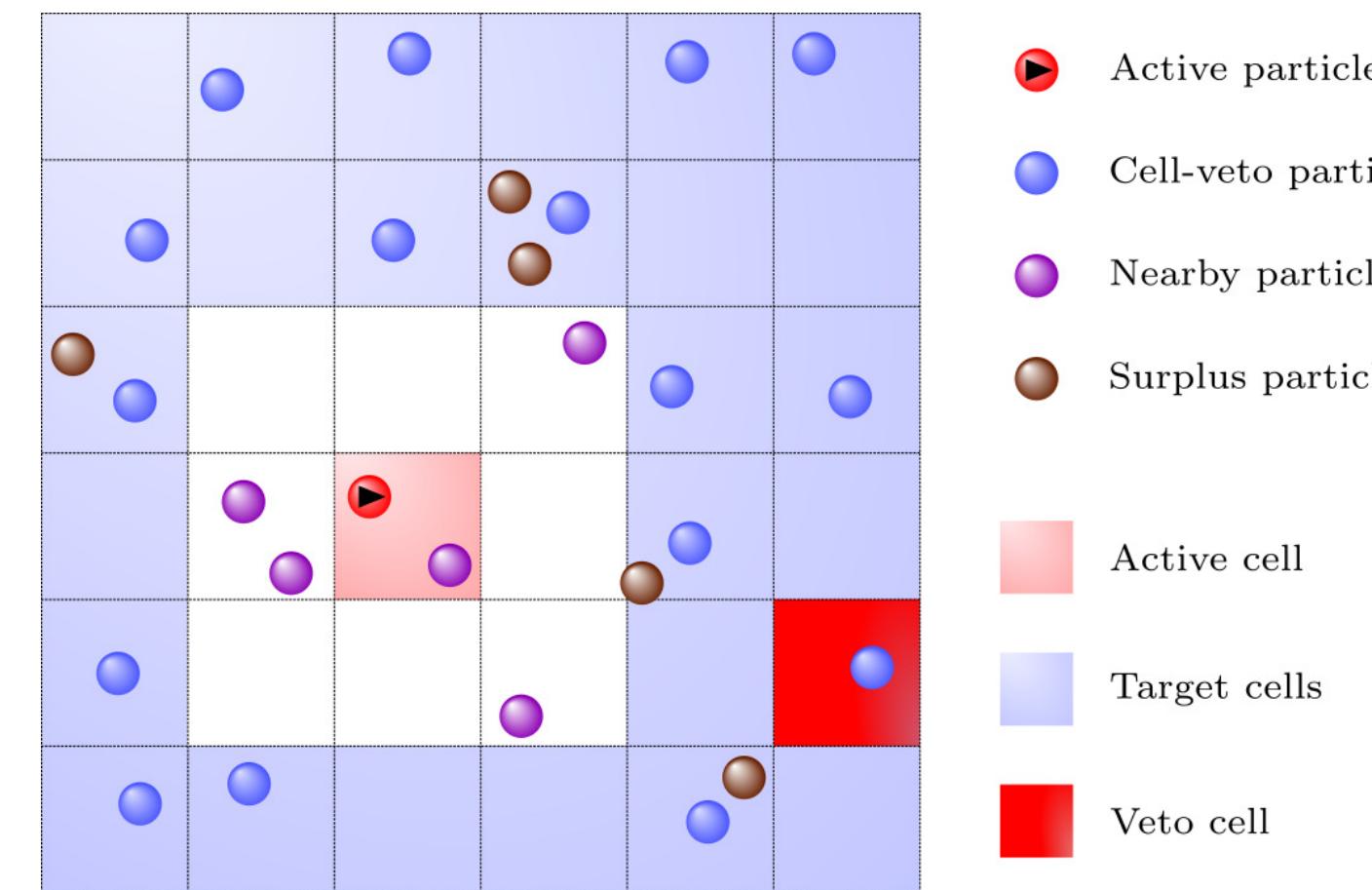
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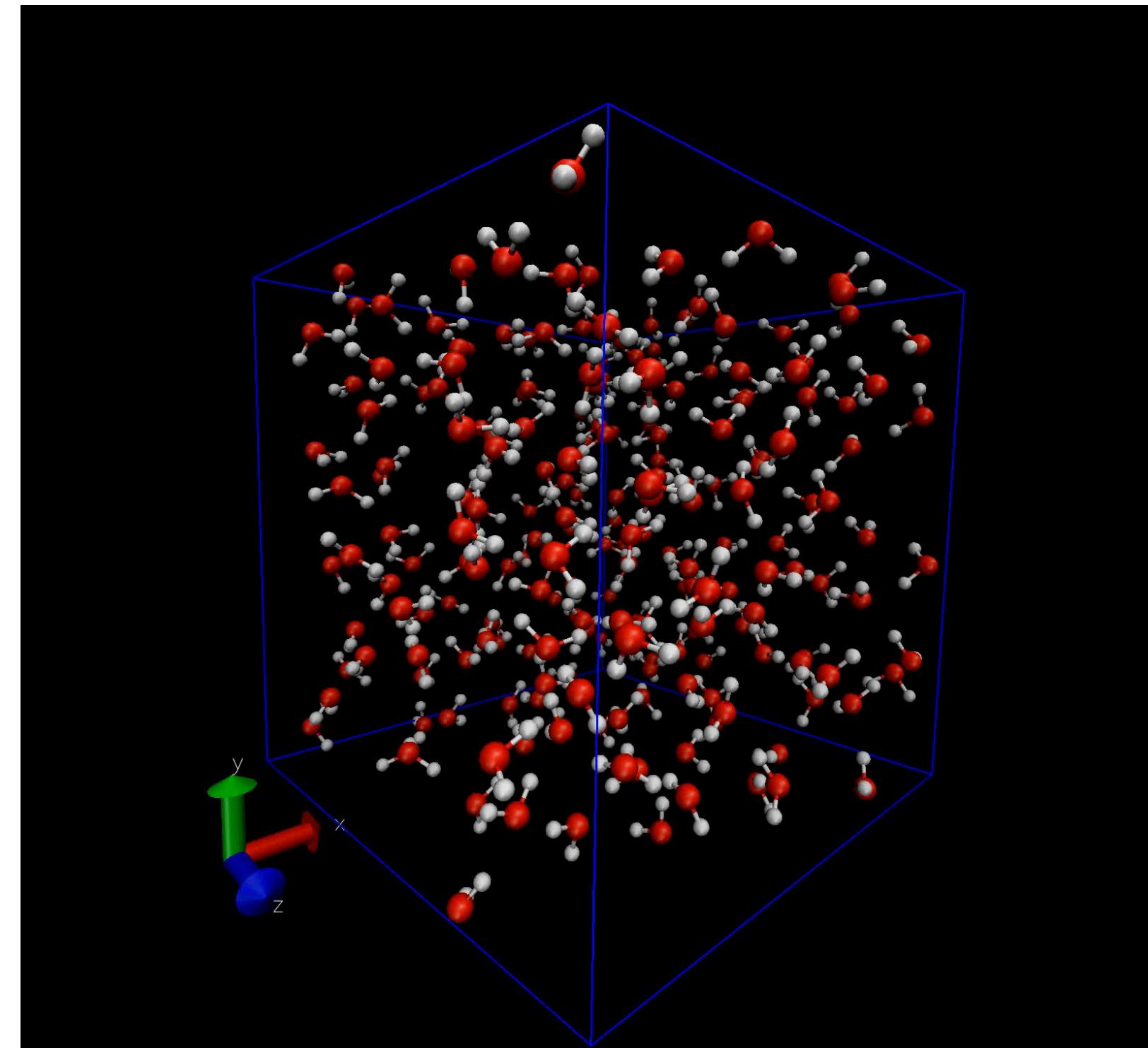
- Cell-veto algorithm considers all non-nearby interactions at once with  $\mathcal{O}(1)$  complexity per event.



[PH](#), All-Atom Event-Chain Monte Carlo – Designing a General-Purpose Python Application (Master's thesis, 2019)

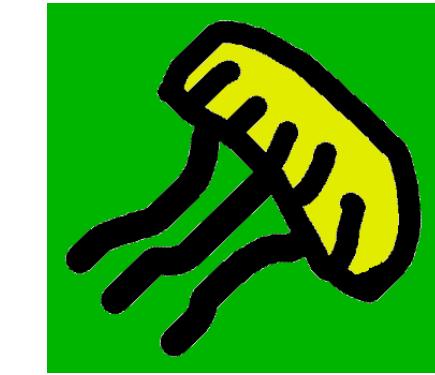
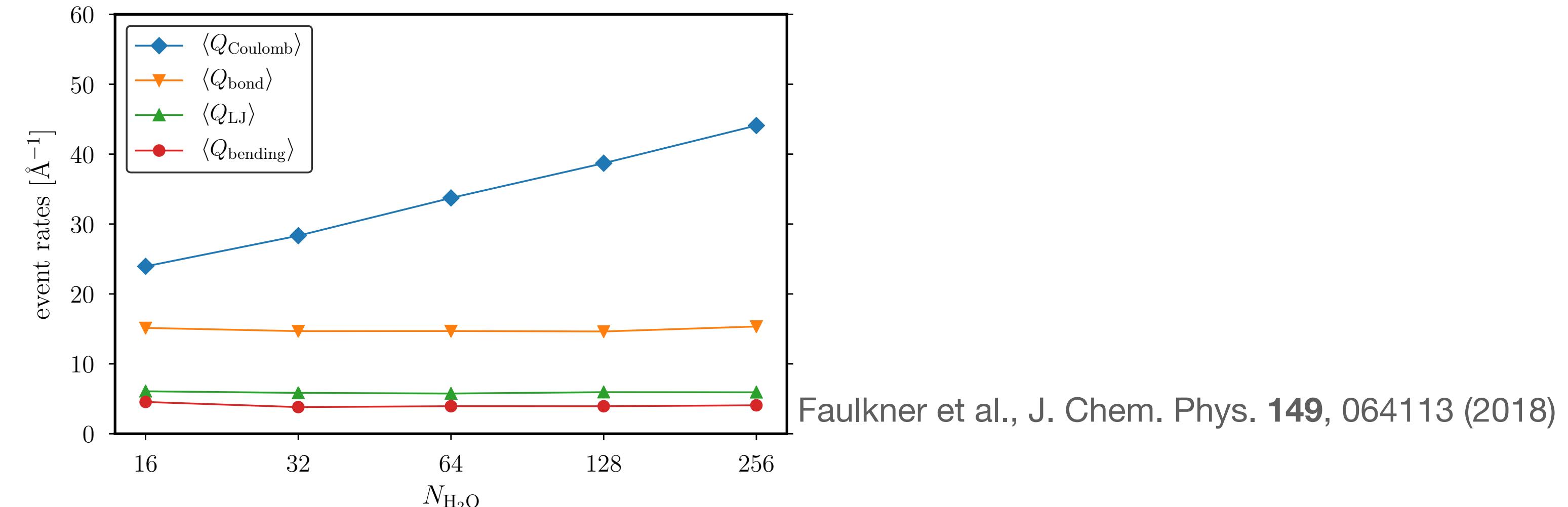
# ECMC for Molecular Systems

- ECMC with cell-veto algorithm samples the Boltzmann distribution  $\exp(-\beta U)\dots$ 
  - ...without ever knowing  $U$ .
  - ...numerically exact without discretizing time (or space).
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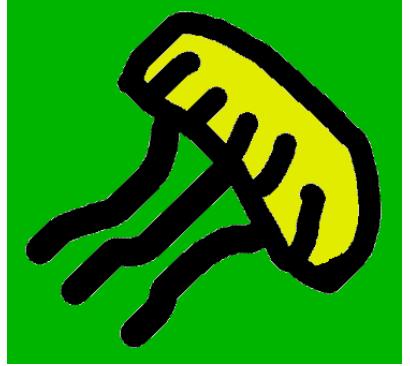


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- Factors involving more than three atoms can be treated in a variety of ways (e.g., have most liftings within a water molecule).
- ECMC advances  $N$  SPC/Fw water molecules with  $\mathcal{O}(N \log N)$  complexity.



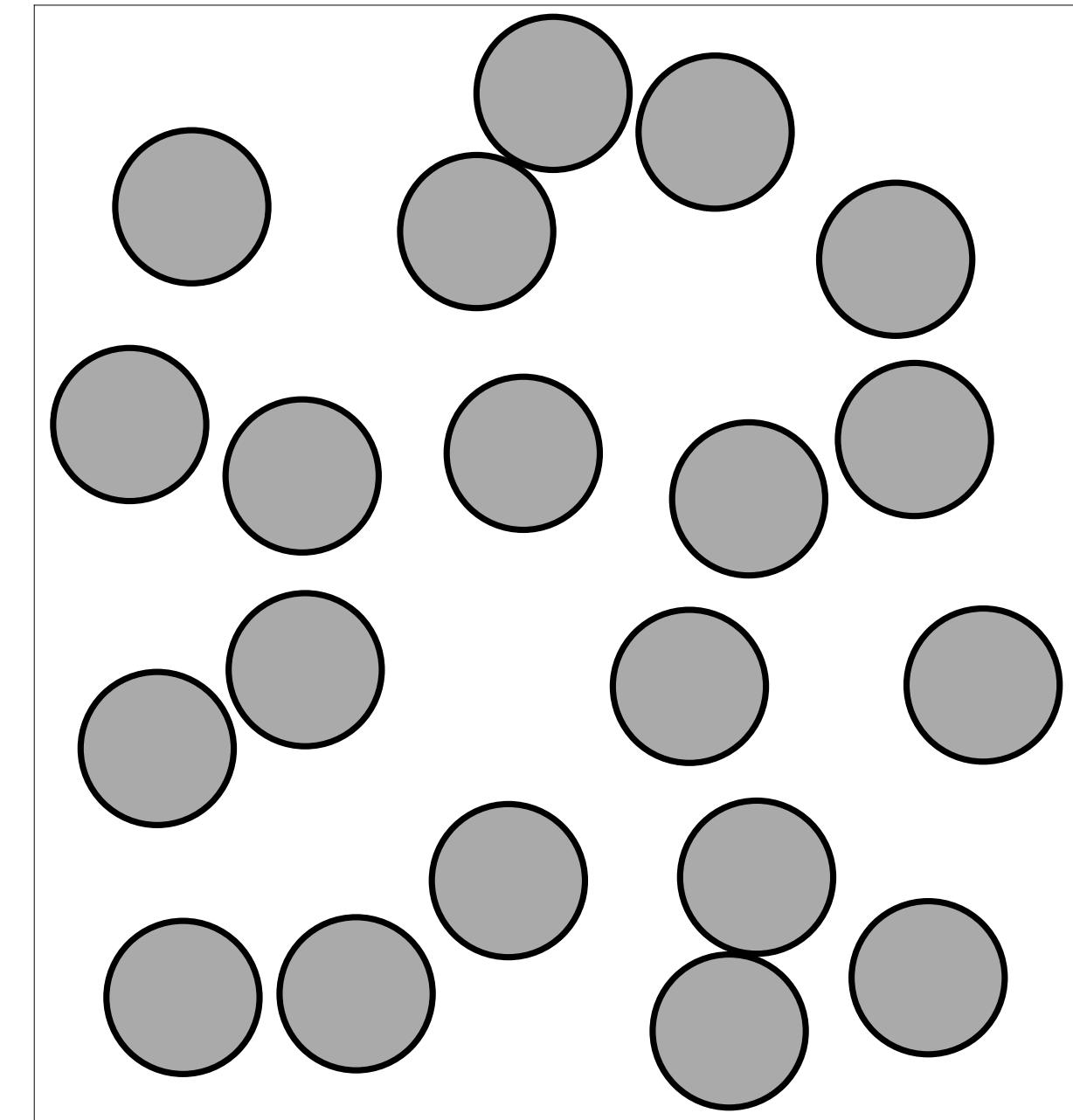
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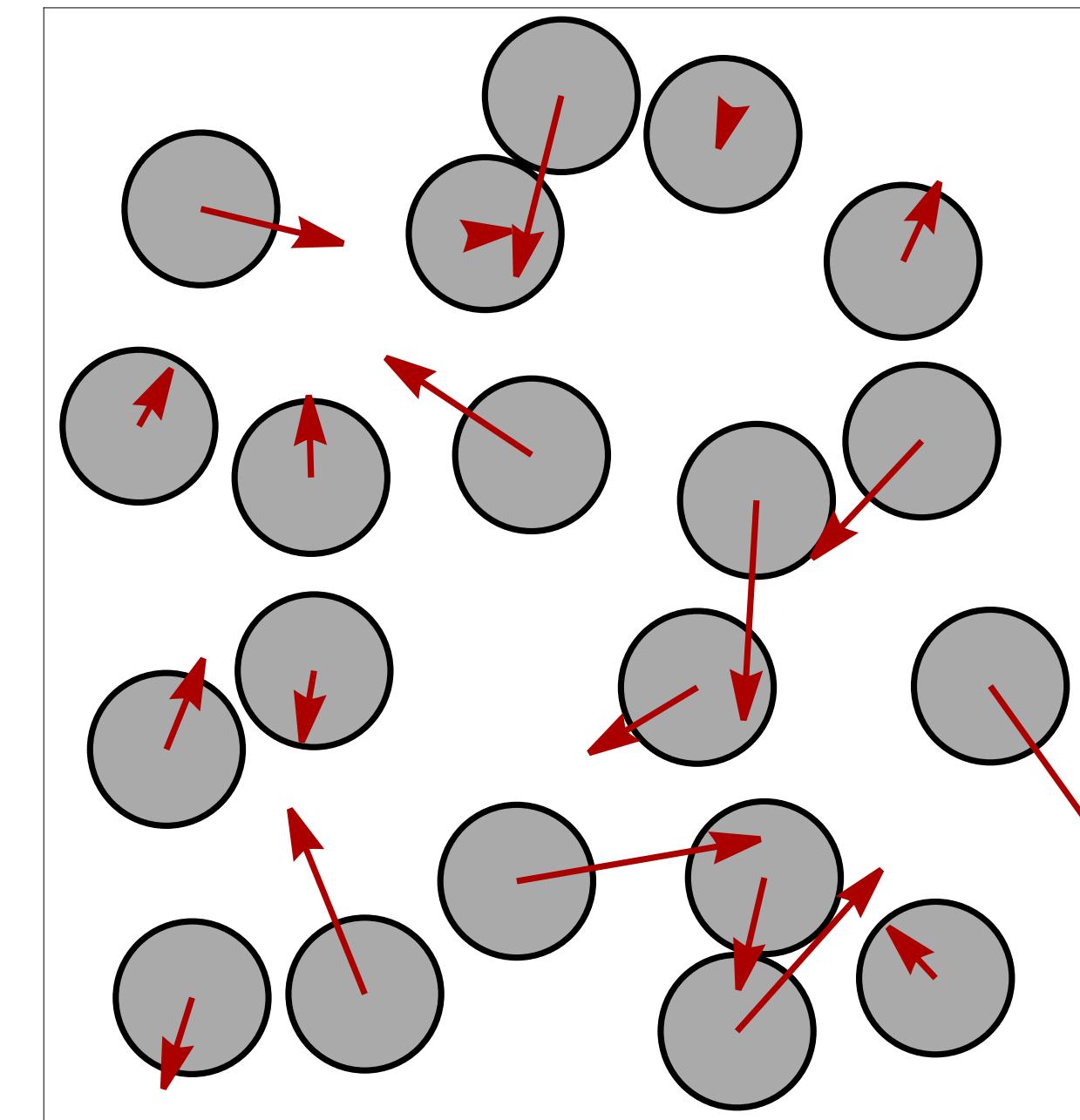
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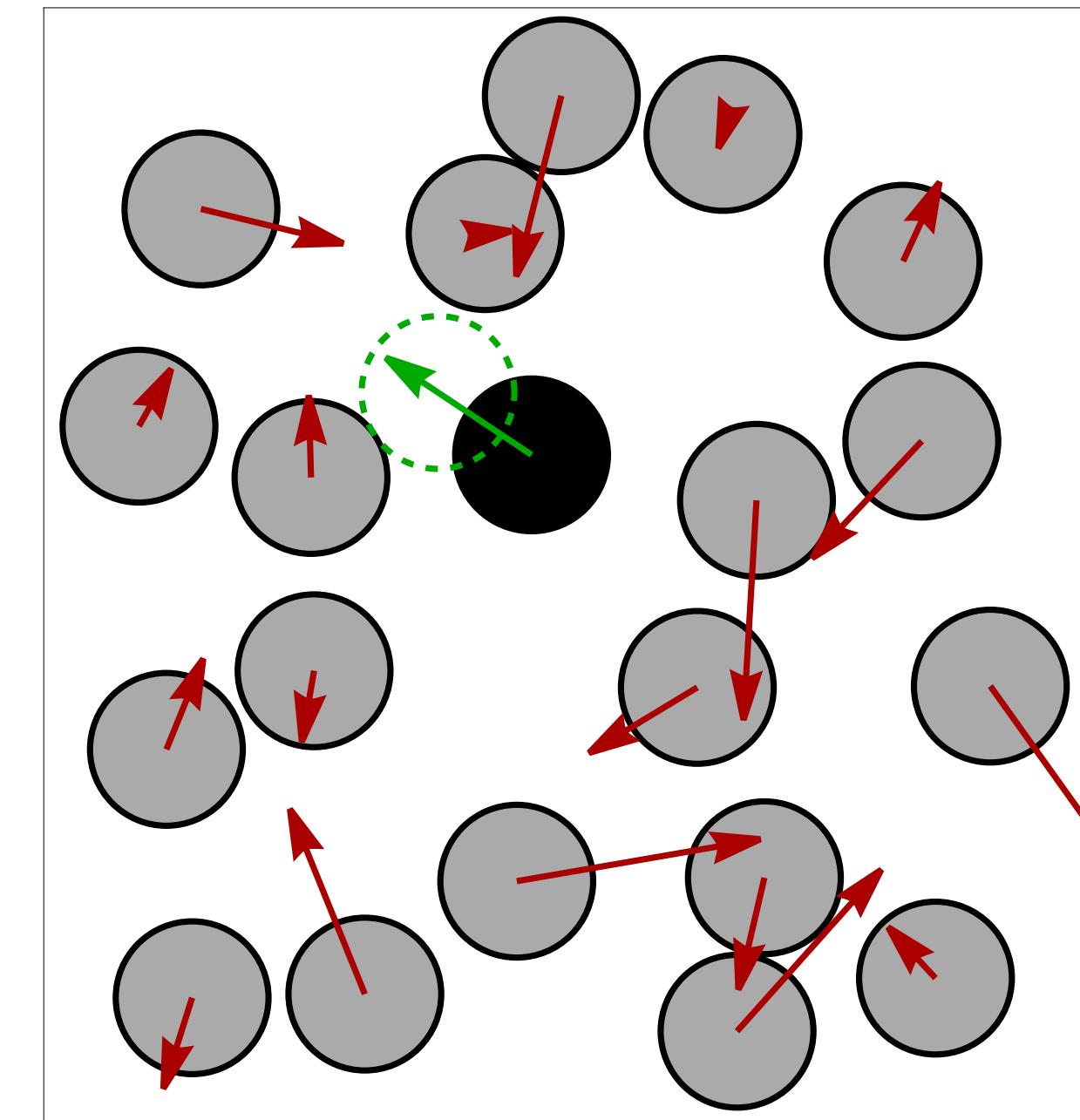
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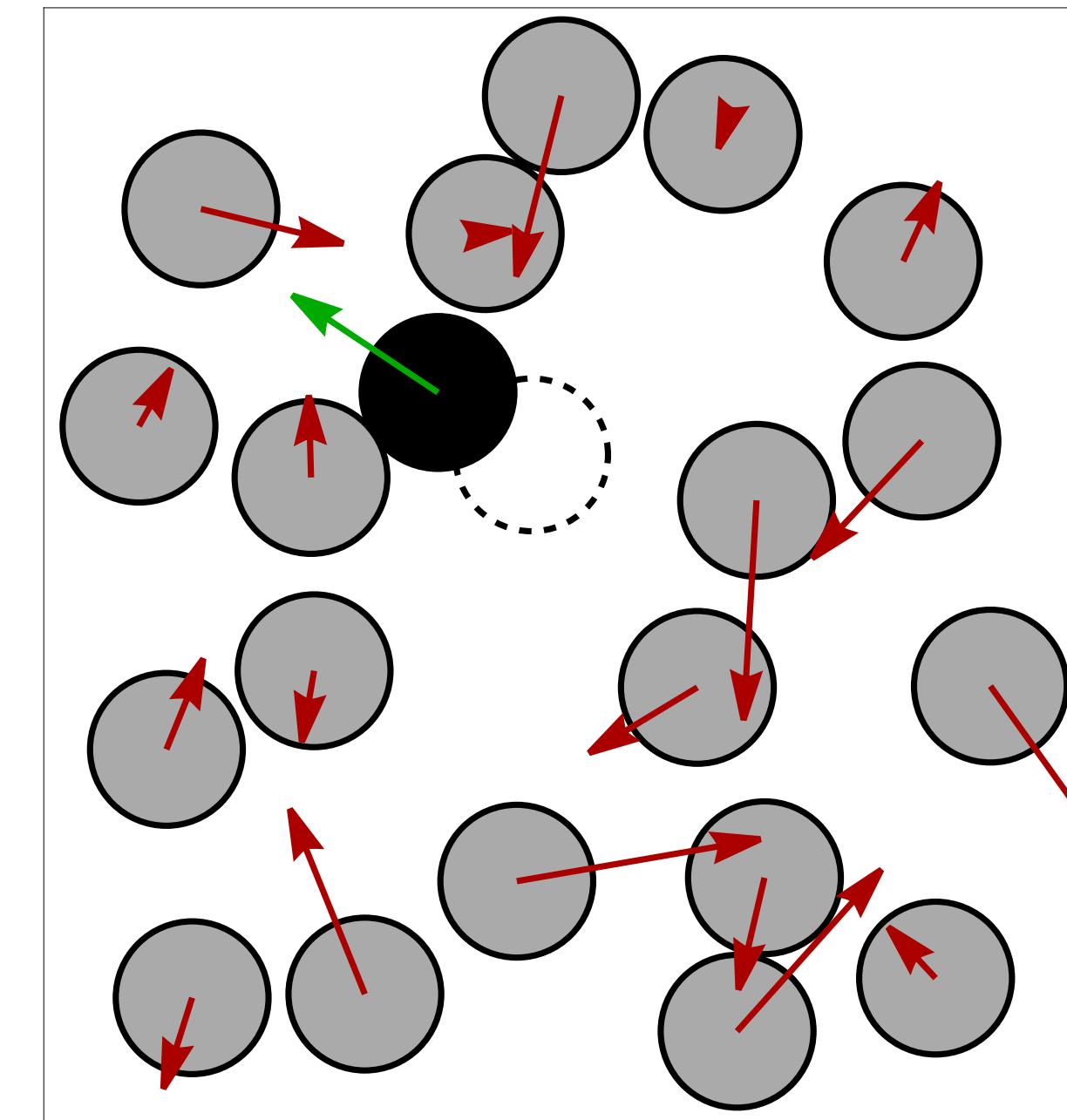
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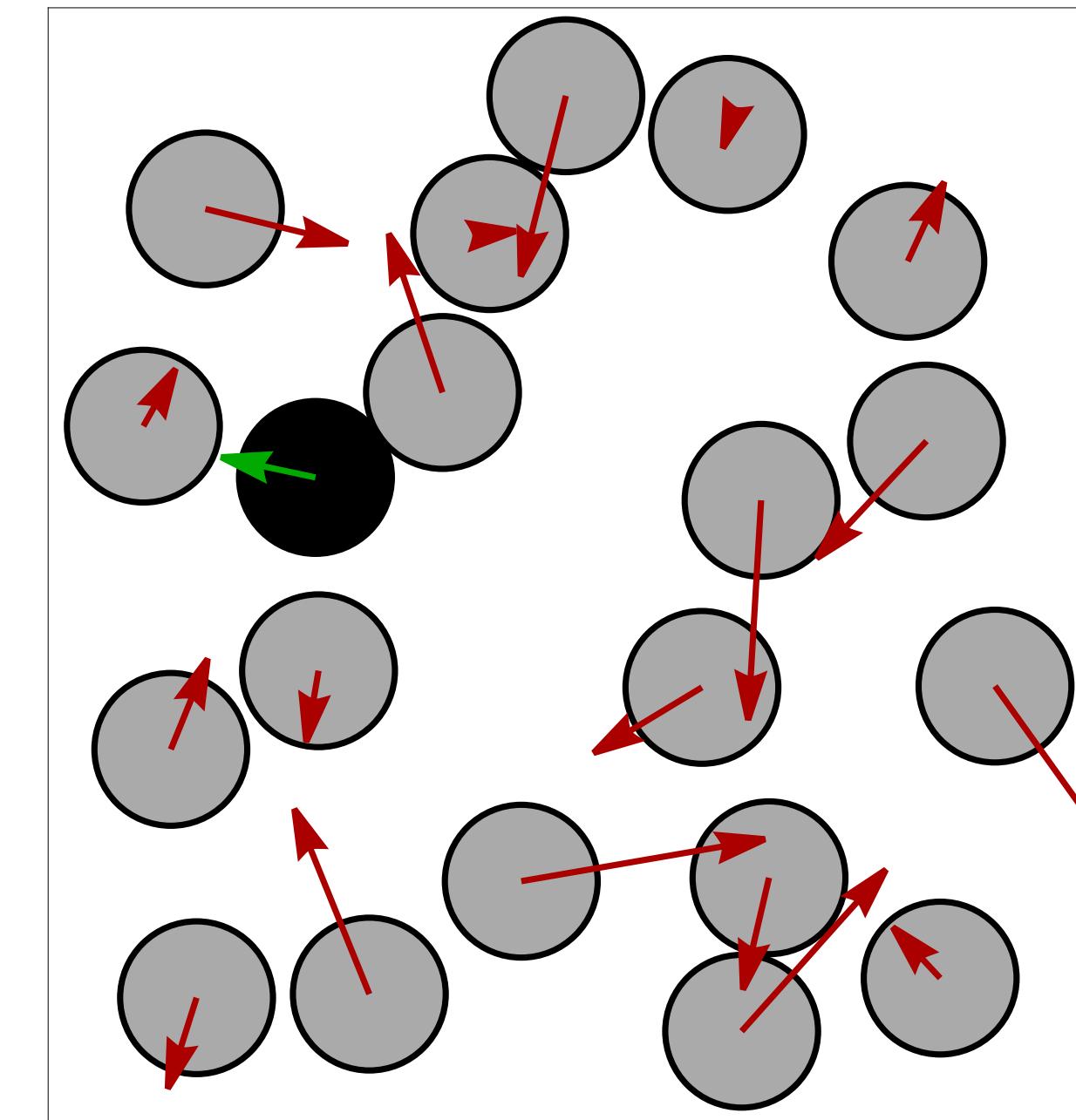
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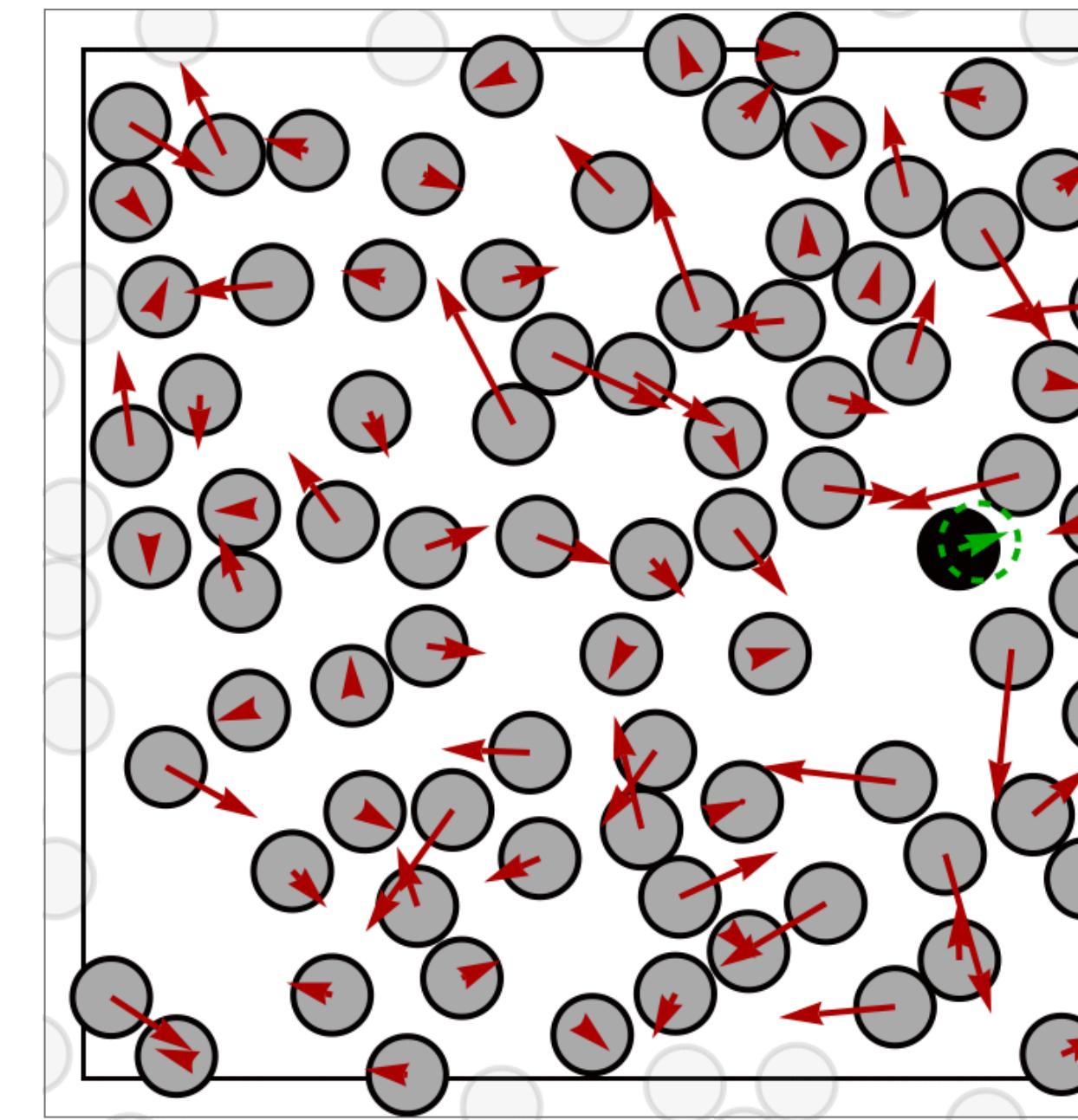
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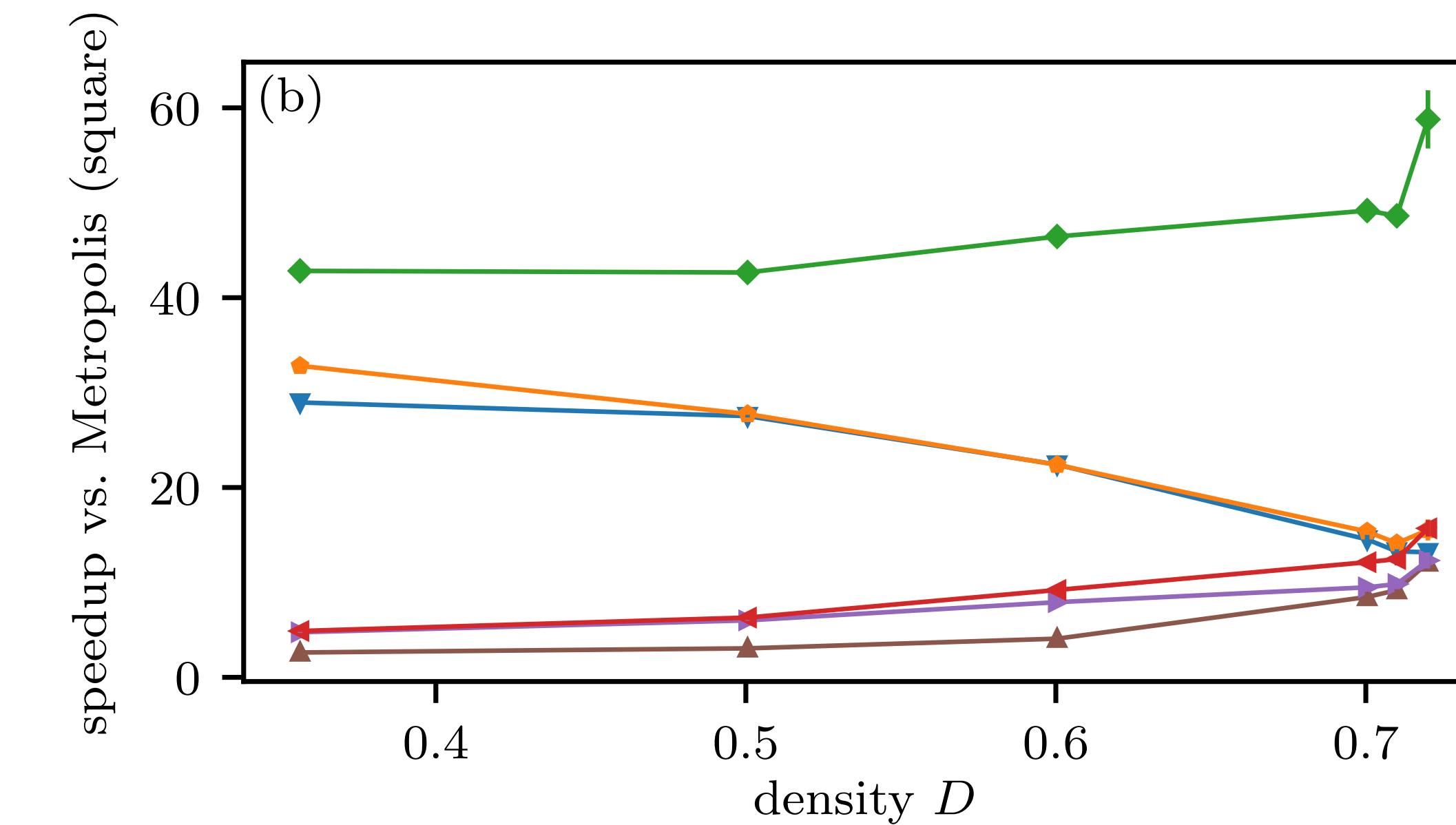
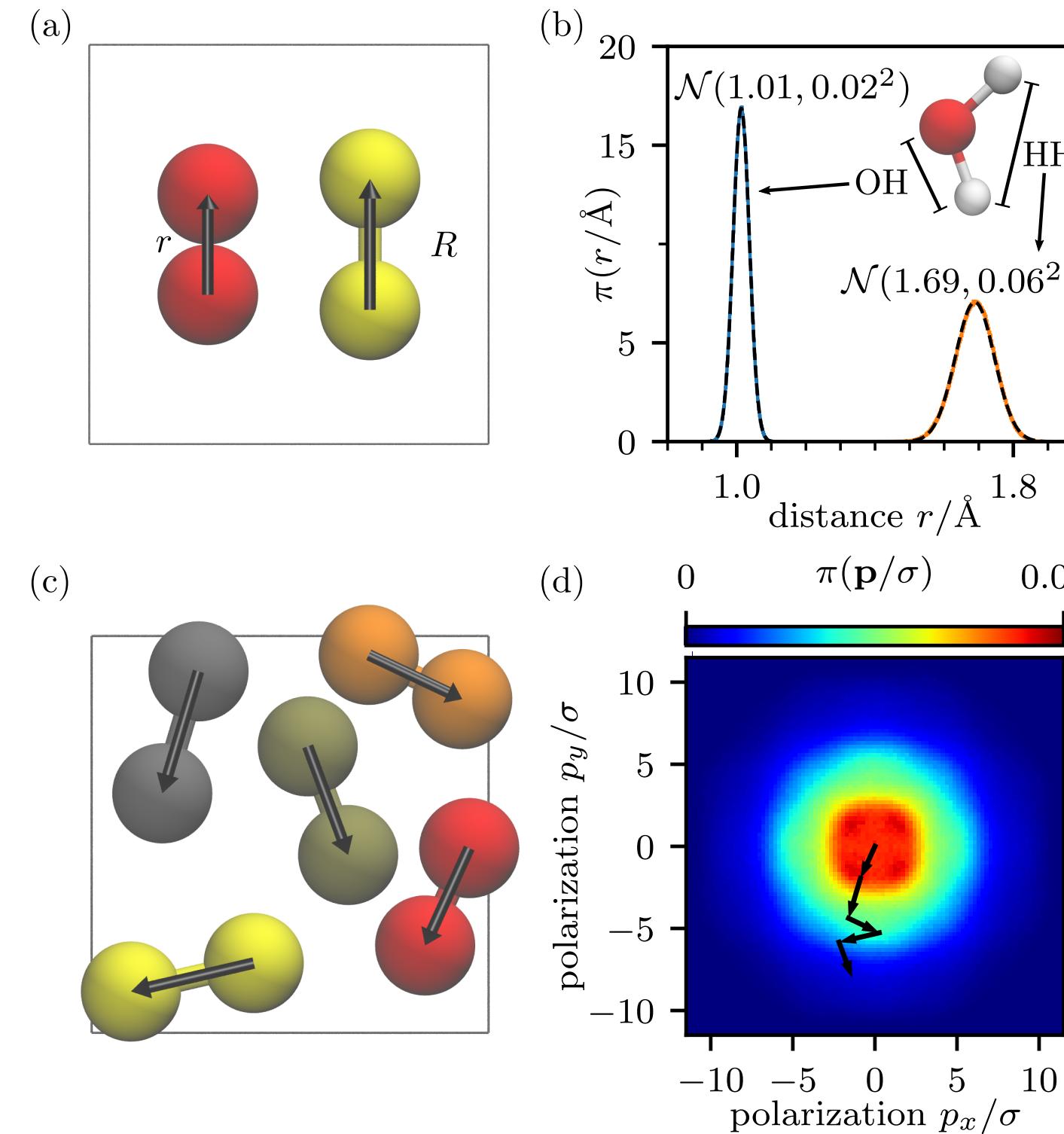
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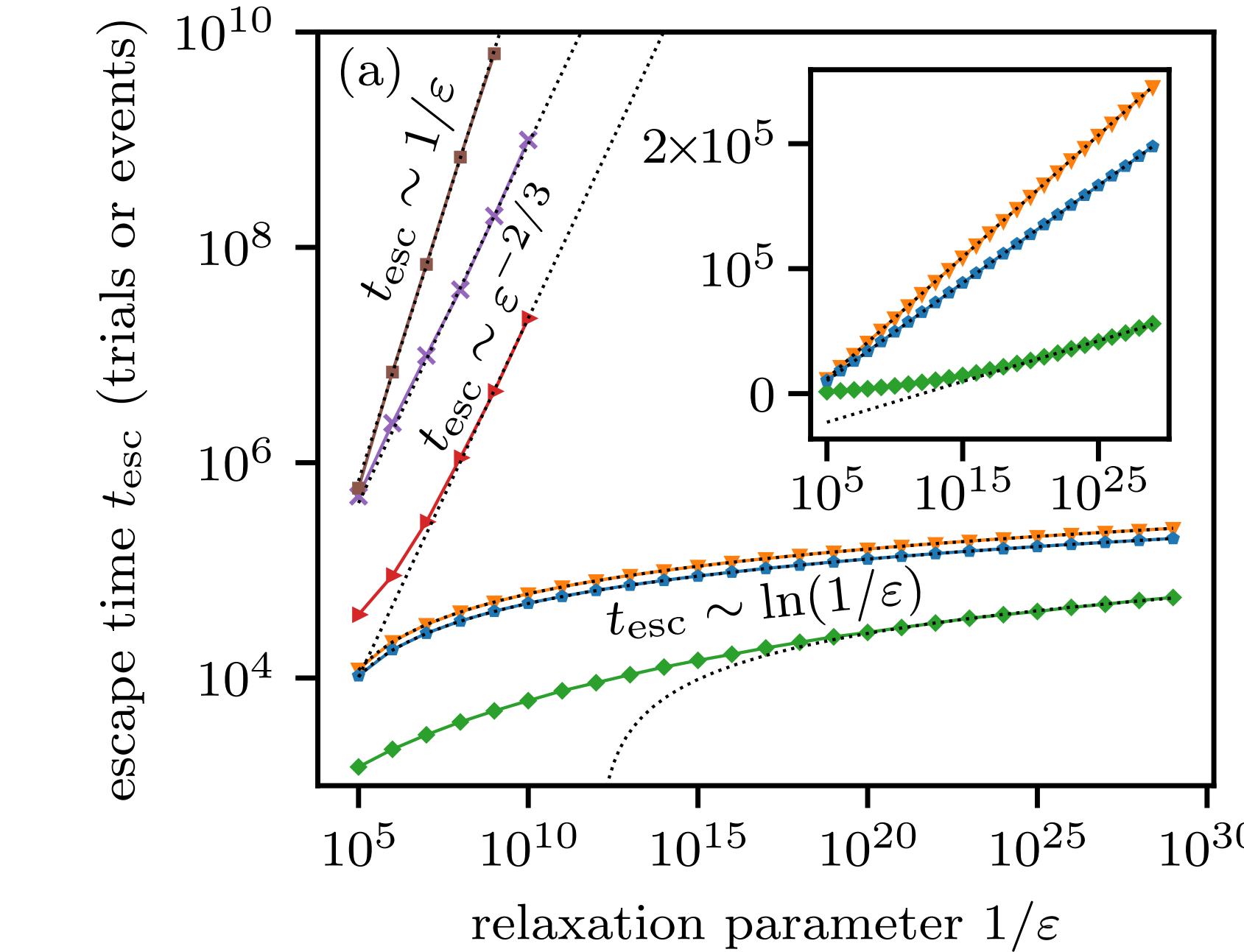
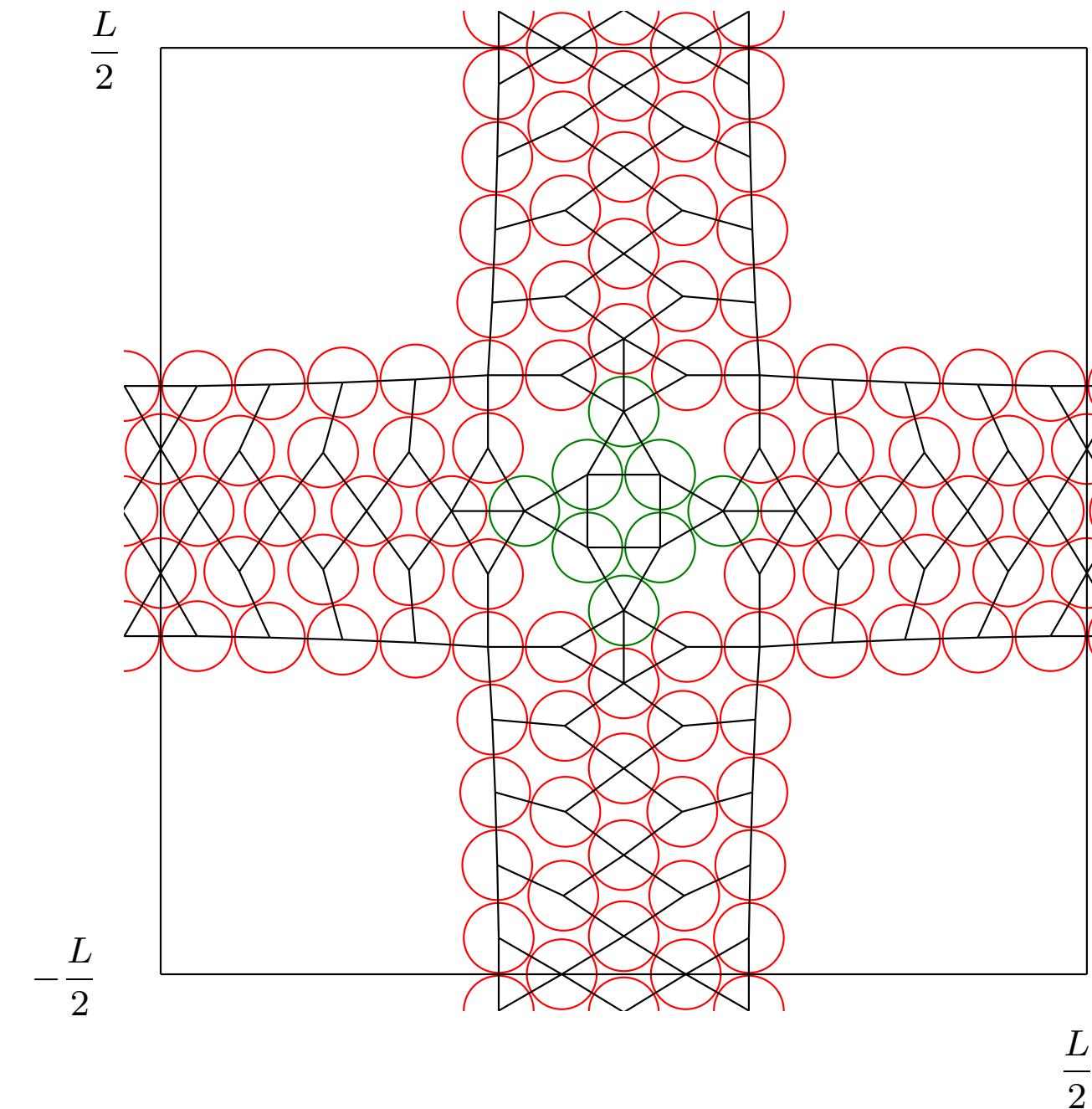
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PH et al., J. Stat. Phys. 187, 31 (2022)

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- Trivial choice of the optimal chain time (very large or even infinite).
  - No fine-tuning of internal parameters required.

# Conclusion

- ECMC with cell-veto algorithm samples the Boltzmann distribution  $\exp(-\beta U)\dots$ 
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# Outlook

- Direct comparison with MD that considers the precision in the MD algorithm.
- Improve dynamics beyond Newtonian dynamics (e.g., with factor fields).
- Efficiently parallelize ECMC.