

Lifted Markov chains: from solvable models to applications in chemical physics

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Equation of State Calculations by Fast Computing Machines

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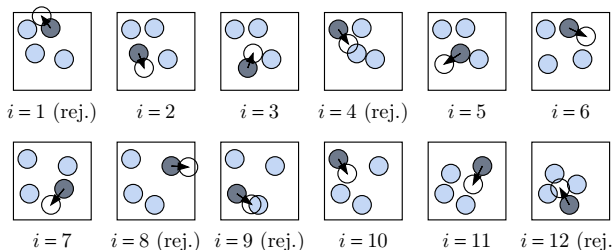
AND

EDWARD TELLER,* *Department of Physics, University of Chicago, Chicago, Illinois*
(Received March 6, 1953)

A general method, suitable for fast computing machines, for investigating such properties as equations of state for substances consisting of interacting individual molecules is described. The method consists of a modified Monte Carlo integration over configuration space. Results for the two-dimensional rigid-sphere system have been obtained on the Los Alamos MANIAC and are presented here. These results are compared to the free volume equation of state and to a four-term virial coefficient expansion.

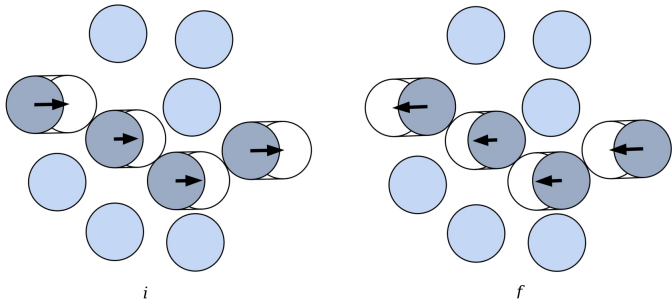


Metropolis algorithm for hard disks



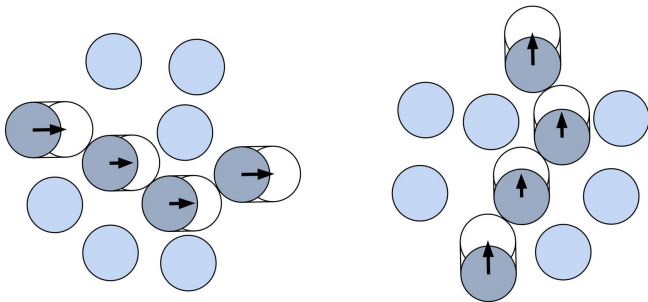
- Metropolis algorithm (1953, mod.)

Event-chain Monte Carlo (reversible)



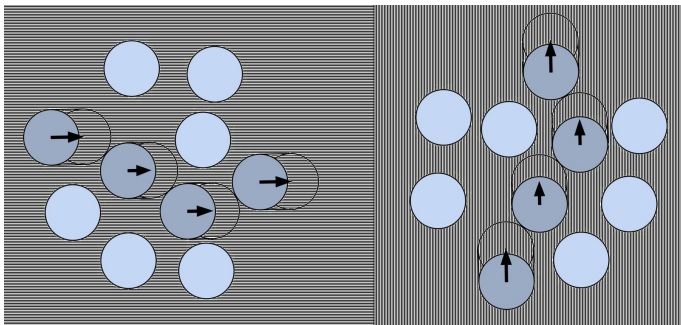
- total length of arrows = ℓ , fixed (sic!)

Event-chain Monte Carlo (non-reversible)



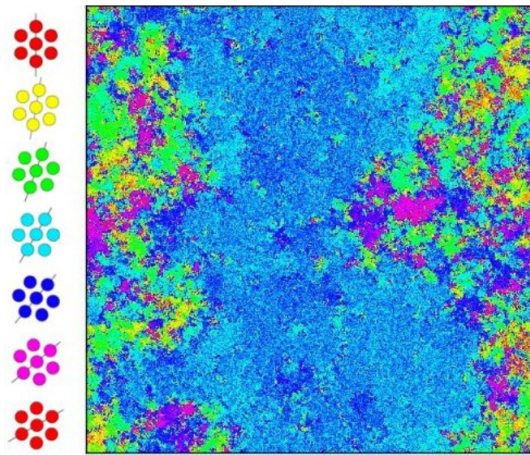
- Box of size $L \times L$, periodic boundary conditions, N disks.
- Event-chain Monte Carlo: Bernard, Krauth, Wilson (2009).

Event-chain Monte Carlo and molecular dynamics



- Box of size $L \times L$, periodic boundary conditions, N disks.
- Event-chain Monte Carlo: Bernard, Krauth, Wilson (2009).
- Molecular dynamics 'on a grating'.
- Hamiltonian Monte Carlo 'on a grating'.

Equilibrium hard-disk configuration



- First equilibrated sample of 10^6 disks ever obtained (Bernard & Krauth 2011, see also Li et al. 2022)
liquid–hexatic coexistence

Markov chains

- **Sample space Ω** (e.g. hard disks, water molecules, quarks, ...)
- **Markov chain** $\leftarrow$ Sequence of random variables
($X_0 \sim \pi^{\{0\}}, X_1 \sim \pi^{\{1\}}, X_2 \sim \pi^{\{2\}} \dots$)
 X_{t+1} depends only on X_t , t is a 'time'
- **Transition matrix P :**
 - P_{ij} : conditional probability to move from sample i to sample j .
 - $\pi^{\{t+1\}} = \pi^{\{t\}} P$: Evolve probability distribution at time t to probability distribution at time $t+1$ (with $\pi^{\{t\}}, t > 0$ often non-explicit, even for $t \rightarrow \infty$).
- **Move set $\mathcal{L}$:** ... from which moves are sampled.
- **Equilibrium distribution π :** Satisfies global balance:

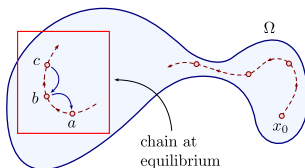
$$\pi_i = \sum_{j \in \Omega} \pi_j P_{ji} \quad \forall i \in \Omega.$$

NB: P irreducible $\implies \pi$ unique.

- **Aperiodicity:** Absence of cycles. P irreducible and aperiodic:

$$\pi^{\{t\}} \rightarrow \pi \quad \text{for } t \rightarrow \infty$$

Detailed balance, global balance



- Reversible transition matrices P satisfy the 'detailed-balance' condition:

$$\pi_a P_{ab} = \pi_b P_{ba}$$

- Non-reversible transition matrices P only satisfy 'global balance':

$$\pi_a = \sum_{b \in \Omega} \pi_b P_{ba}$$

Conductance (bottleneck ratio)

$$\Phi \equiv \min_{S \subset \Omega, \pi_S \leq \frac{1}{2}} \frac{\mathcal{F}_{S \rightarrow \bar{S}}}{\pi_S} = \min_{S \subset \Omega, \pi_S \leq \frac{1}{2}} \frac{\sum_{i \in S, j \notin S} \pi_i P_{ij}}{\pi_S}.$$

- Reversible Markov chains:

$$\frac{1}{\Phi} \leq \tau_{\text{rel}} \leq \frac{8}{\Phi^2}$$

($\leq$: Sinclair & Jerrum (1986), Lemma (3.3))

- Arbitrary Markov chain (see Chen et al. (1999)):

$$\frac{1}{4\Phi} \leq \mathcal{A} \leq \frac{20}{\Phi^2},$$

($\mathcal{A}$: set time: Expectation of $\max_S (t_S \times \pi_S)$ from equilibrium)

- Markov chain $\Pi (\Omega, P, \pi) \Leftrightarrow$ Lifted Markov chain $\hat{\Pi} (\hat{\Omega}, \hat{P}, \hat{\pi})$
- Mapping f from $\hat{\Omega}$ to Ω .
- **Condition 1:** π is preserved

$$\pi_v = \hat{\pi} [f^{-1}(v)] = \sum_{i \in f^{-1}(v)} \hat{\pi}_i,$$

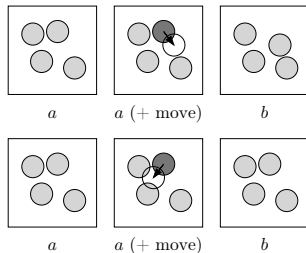
- **Condition 2:** flows are preserved

$$\underbrace{\pi_v P_{vu}}_{\text{collapsed flow}} = \sum_{i \in f^{-1}(v), j \in f^{-1}(u)} \overbrace{\hat{\pi}_i \hat{P}_{ij}}^{\text{lifted flow}}.$$

- $\hat{P}$ cannot have larger (better) conductance than P .
- Our work: $\hat{\Omega} = \Omega \times \bar{\mathcal{L}}$ with $\bar{\mathcal{L}}$ subset of move set $\mathcal{L}$.
- Speedup requires non-reversibility (Chen et al. 1999).

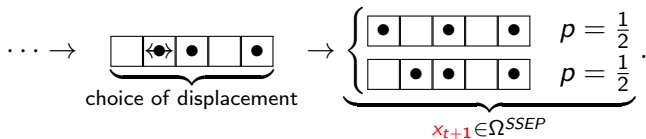
The Metropolis algorithm and the SSEP

1 Metropolis (1953) algorithm:



The Metropolis algorithm is reversible.

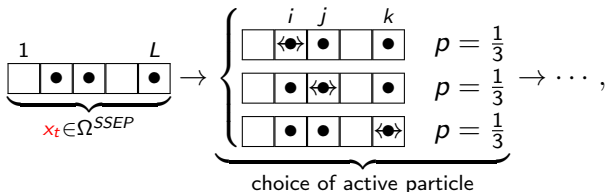
2 SSEP:



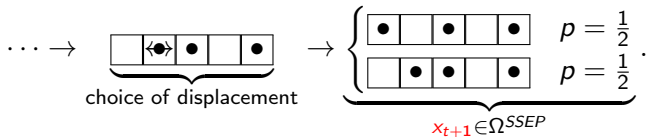
1D lattice reduction of hard-sphere Metropolis algorithm.

Symmetric simple exclusion process (SSEP)

- Move (first part ...)



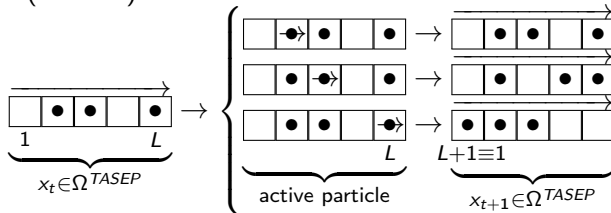
- Move (... second part)



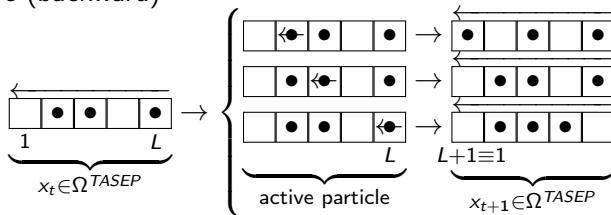
- The SSEP is 'our' collapsed (reversible) Markov chain.
- $\pi = \text{const.}$

Totally asymmetric simple exclusion process (TASEP)

- Move (forward)



- Move (backward)



- forward-backward coupling (ad-hoc, or boundary conditions).

NB: Non-reversible, i.e. non-equilibrium, but samples equilibrium Boltzmann distribution.

Lifted TASEP (definition)

- $\Omega^{l-TASEP} = \Omega^{SSEP} \times \{-1, +1\} \times \{1, \dots, N\}$, $\mathcal{L} = \emptyset$
- Move (first part ...)

$$\underbrace{\begin{array}{|c|c|c|c|} \hline \square & \bullet & \bullet \rightarrow & \square \\ \hline \end{array}}_{x_t \in \Omega^{l-TASEP}} \rightarrow \underbrace{\begin{array}{|c|c|c|c|} \hline \square & \bullet & \square & \bullet \rightarrow \\ \hline \end{array}}_{\text{'particle' displacement}} \rightarrow \dots \quad (1)$$

$$\underbrace{\begin{array}{|c|c|c|c|} \hline \square & \bullet & \bullet \rightarrow & \bullet \\ \hline \end{array}}_{x_t \in \Omega^{l-TASEP}} \rightarrow \underbrace{\begin{array}{|c|c|c|c|} \hline \square & \bullet & \bullet & \bullet \rightarrow \\ \hline \end{array}}_{\text{'lifting' move}} \rightarrow \dots \quad (2)$$

- Move (second part ...)

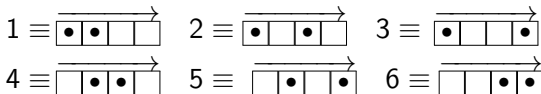
$$\dots \rightarrow \begin{array}{|c|c|c|c|} \hline \square & \bullet & \square & \bullet \rightarrow \\ \hline \end{array} \rightarrow \begin{cases} \begin{array}{|c|c|c|c|} \hline \square & \bullet \rightarrow & \square & \bullet \\ \hline \end{array} & p = \alpha \\ \begin{array}{|c|c|c|c|} \hline \square & \bullet & \square & \bullet \rightarrow \\ \hline \end{array} & p = \beta \end{cases} \quad (2b)$$

$$\dots \rightarrow \begin{array}{|c|c|c|c|} \hline \square & \bullet & \bullet & \bullet \rightarrow \\ \hline \end{array} \rightarrow \begin{cases} \begin{array}{|c|c|c|c|} \hline \square & \bullet & \bullet \rightarrow & \bullet \\ \hline \end{array} & p = \alpha \\ \begin{array}{|c|c|c|c|} \hline \square & \bullet & \bullet & \bullet \rightarrow \\ \hline \end{array} & p = \beta \end{cases} \quad (2b)$$

$x_{t+1} \in \Omega^{lTASEP}$

TASEP (example)

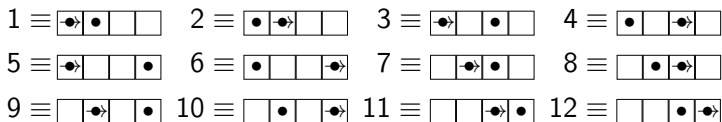
NB: Consider only the forward-moving sector (pbc):



$$P^{\text{TASEP}} = \frac{1}{2} \begin{bmatrix} 1 & 1 & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & 1 & 1 & \cdot & \cdot \\ \cdot & \cdot & 1 & \cdot & 1 & \cdot \\ \cdot & \cdot & \cdot & 1 & 1 & \cdot \\ 1 & \cdot & \cdot & \cdot & \cdot & 1 \\ \cdot & 1 & \cdot & \cdot & \cdot & 1 \end{bmatrix} .$$

Lifted TASEP (example)

NB: Consider only the forward-moving sector (pbc):



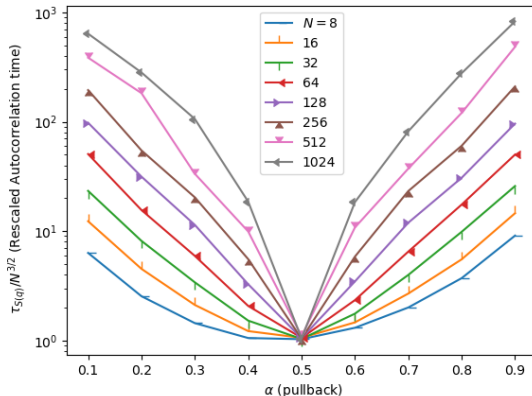
$$P = \begin{bmatrix} \boxed{\begin{smallmatrix} \alpha & \beta \\ \cdot & \cdot \end{smallmatrix}} & \boxed{\begin{smallmatrix} \cdot & \cdot \\ \alpha & \beta \end{smallmatrix}} & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \boxed{\begin{smallmatrix} \cdot & \cdot \\ \alpha & \beta \end{smallmatrix}} & \boxed{\begin{smallmatrix} \beta & \alpha \\ \cdot & \cdot \end{smallmatrix}} & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \boxed{\begin{smallmatrix} \cdot & \cdot \\ \beta & \alpha \end{smallmatrix}} & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \boxed{\begin{smallmatrix} \cdot & \cdot \\ \beta & \alpha \end{smallmatrix}} & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \end{bmatrix}$$

Synopsis of mixing and relaxation times

Algorithm	mixing	inverse gap	Comment
SSEP	$N^3 \log N$	N^3	Edwards–Wilkinson
TASEP	$N^{5/2}$	$N^{5/2}$	KPZ
Lifted TASEP	N^2	$N^2 (N^{3/2})$	at α_{crit} only, otherwise KPZ.

- continuous-space versions available (Kapfer & Krauth (2017)).
- see Essler & Krauth (2023), results from Bethe ansatz.

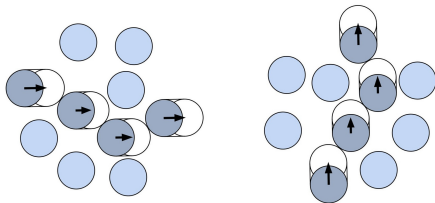
Lifted TASEP benchmark



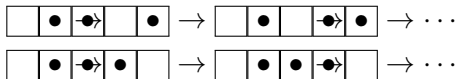
- Autocorrelations of structure factor (small- q density fluctuations) of the lifted TASEP.
- $\alpha = \alpha_{\text{crit}}$: vanishing drift velocity of activity.

The lifted TASEP and event-chain Monte Carlo

- 1 Event-chain Monte Carlo algorithm (Bernard, Krauth, Wilson 2009):



- 2 Lifted TASEP:



One-dimensional lattice reduction of event-chain Monte Carlo.

Factorized Metropolis algorithm (pair potential)

- Metropolis filter:

$$p^{\text{Met}}(a \rightarrow b) = \min \left[1, \exp \left(-\beta \sum_{i < j} \Delta U_{i,j} \right) \right]$$

$$p^{\text{Met}}(a \rightarrow b) = \min \left[1, \prod_{i < j} \exp(-\beta \Delta U_{i,j}) \right]$$

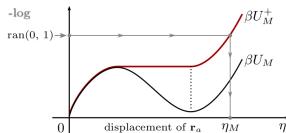
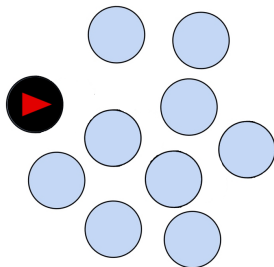
- Factorized Metropolis filter (Michel, Kapfer, Krauth 2014):

$$p^{\text{Fact}}(a \rightarrow b) = \prod_{i < j} \min [1, \exp(-\beta \Delta U_{i,j})].$$

$$X^{\text{Fact}}(a \rightarrow b) = X_{1,2} \wedge X_{1,3} \wedge \cdots \wedge X_{N-1,N}$$

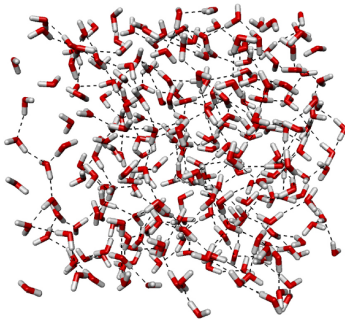
- i.e.: Accept move if all pairs (i,j) accept it.
- **Consensus**-based, not gradient-based.

Factorized Metropolis algorithm (pair potential)

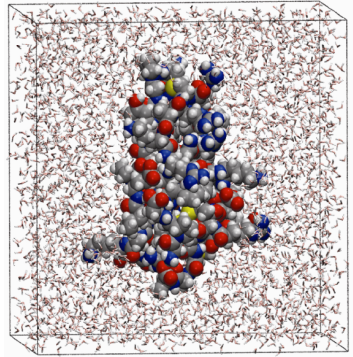


- Complexity $\mathcal{O}(1)$ per event for $N \rightarrow \infty$ (Kapfer–Krauth 2016)
- Zoo of choice about what to do after the event.
- Piecewise deterministic Markov process, samples $\exp(-\beta U)$ at all continuous times.

All-atom simulation with non-reversible Markov chains



(a)

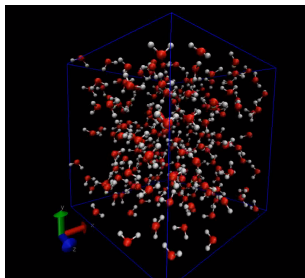


(b)

(a): Höllmer, Maggs, Krauth (2023) (see also Faulkner et al. (2018), Höllmer et al. (2020)).

(b): Yet to be done...

All-Atom Coulomb problem

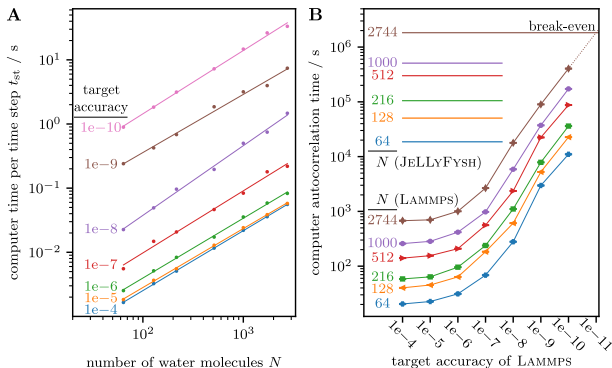


- Factors M : 'bond', 'bending', 'Lennard-Jones', 'Coulomb' (SPC/Fw water model).

$$X^{\text{Fact}}(c \rightarrow c') = \bigwedge_{M \in \mathcal{M}} X_M(c_M \rightarrow c'_M).$$

- i.e.: Move until a single of $\mathcal{O}(N)$ factors vetoes acceptance.
- **Complexity $\mathcal{O}(1)$ per event** (Kapfer, Krauth 2016).
- Event-driven, approximation-free, canonic, without thermostat.

Water simulation—benchmark JeLLyFysh—LAMMPS



- See Höllmer, Maggs, Krauth (2023)
- JeLLyFysh (non-reversible Markov chain) is approximation-free: Evaluates neither forces nor total potentials.
- LAMMPS (molecular dynamics) has time-step error and lattice error for computation of electrostatic interaction.

Lifted Markov chains:

- From the beginnings to real-world applications.
- Event-driven, rejection-free, non-reversible, fast.
- Consensus-driven, not gradient-driven.
- Non-reversibility at all times appears essential.

Outlook:

- Sampling $\exp(-\beta U)$ without evaluating U .
- 'Natively cutoff-free' MCMC (Coulomb, LJ) in $\mathcal{O}(1)$.
- Applications in chemical physics.