Mixing, stopping, lifting, and other keys to the second Markov-chain revolution

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CMT Forum: Department of Physics, University of Oxford W. Krauth; Front. Phys. (2021) (review)

G. Tartero, W. Krauth; AJP to appear (2023) (1+12 miniature algorithms)
 P. Höllmer, A. C. Maggs, W. Krauth (2023) (Real-life application)

F. H. L. Essler, W. Krauth (2023) (Bethe ansatz) Z. Lei, W. Krauth, A. C. Maggs; PRE (2019) (Factor fields) M. Michel, S. Kapfer, W. Krauth; JCP (2014) (Factorization)



Introduction

- Ø Mixing and Relaxing, stopping
- 3 Factoring, balance-breaking
- Moving towards real-life problems
- Onclusion



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Markov chains

- Sample space Ω (e.g. point particles, water molecules)
- 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 $t \to \infty$).
- 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:

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

Total variation distance, mixing time

• Total variation distance:

$$||\pi^{\{t\}} - \pi||_{\mathsf{TV}} = \max_{A \subset \Omega} |\pi^{\{t\}}(A) - \pi(A)| = \frac{1}{2} \sum_{i \in \Omega} |\pi_i^{\{t\}} - \pi_i|.$$

Distance:

$$d(t) = \max_{\pi^{\{0\}}} ||\pi^{\{t\}}(\pi^{\{0\}}) - \pi||_{\mathsf{TV}}$$

• Mixing time:

$$t_{\min}(\epsilon) = \min\{t : d(t) \le \epsilon\}$$

• Usually $\epsilon=1/4$ is taken, $\epsilon=1/e$ would be better.



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Shuffling of cards 1/6



• $\Omega_N^{\text{shuffle}} = \{\text{Permutations of } \{1, \dots, N\}\}$

• For
$$N = 3$$
:
 $\Omega_3^{\text{shuffle}} = \{1 \equiv \{1, 2, 3\}, 2 \equiv \{1, 3, 2\}, 3 \equiv \{2, 1, 3\}, 4 \equiv \{2, 3, 1\}, 5 \equiv \{3, 1, 2\}, 6 \equiv \{3, 2, 1\}\}.$
• $\pi^{t=0} = \delta((1, ..., N))$



Shuffling of cards 2/6



moves

procedure top-to-random input $\{c_1, \ldots, c_n\}$ $i \leftarrow \text{choice}(\{1, \ldots, n\})$ $\{\hat{c}_1, \ldots, \hat{c}_n\} \leftarrow \{c_2, \ldots, c_i, c_1, c_{i+1}, \ldots, c_n\}$ output $\{\hat{c}_1, \ldots, \hat{c}_n\}$

- Insert upper card (c_1) after card i and before card i + 1
- NB: if i = 1, put it back on top.



Shuffling of cards 3/6

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moves

•
$$\Omega_3^{shuffle} = \{1 \equiv \{1, 2, 3\}, 2 \equiv \{1, 3, 2\}, 3 \equiv \{2, 1, 3\}, 4 \equiv \{2, 3, 1\}, 5 \equiv \{3, 1, 2\}, 6 \equiv \{3, 2, 1\}\}.$$

$$P = \frac{1}{3} \begin{pmatrix} 1 & 0 & 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 1 & 1 \\ 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 & 0 & 1 \end{pmatrix}$$



Shuffling of cards 4/6

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$$P_{3}^{\mathsf{shuffle}} = \frac{1}{3} \begin{pmatrix} 1 & 0 & 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 1 & 1 \\ 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 & 0 & 1 \end{pmatrix}$$

• Eigenvalues of P_N^{shuffle} : $0, \frac{1}{N}, \frac{2}{N}, \dots, 1 - \frac{2}{N}, 1$

• Degeneracies:

$$N = 2 : [1, 0, 1]$$

$$N = 3 : [2, 3, 0, 1]$$

$$N = 4 : [9, 8, 6, 0, 1]$$

$$N = 5 : [44, 45, 20, 10, 0, 1]$$

$$N = 6 : [265, 264, 135, 40, 15, 0, 1]$$

$$N = 7 : [1854, 1855, 924, 315, 70, 21, 0, 1]$$



Shuffling of cards 5/6



moves

 $\begin{array}{l} \textbf{procedure top2random-stop} \\ \textbf{input} \{c_1, \dots, c_n\} \\ c_{\text{first-n}} \leftarrow c_n \\ \textbf{for } t = 1, 2, \dots \ \textbf{do} \\ \left\{ \begin{array}{l} \tilde{c}_1 \leftarrow c_1 \\ \{c_1, \dots, c_n\} \leftarrow \texttt{top2random}(\{c_1, \dots, c_n\}) \\ \textbf{if } (\tilde{c}_1 = c_{\text{first-n}}) \ \textbf{break} \\ \textbf{output } \{c_1, \dots, c_n, t\} \end{array} \right. \end{array}$

• Expected running time: $n \log n$.



Shuffling of cards 6/6



• Time scale N log N larger than inverse gap.



Mixing and Relaxation



•
$$t_{\text{mix}} = ||\pi^{\{t_{\text{mix}}\}} - \pi||_{\text{TV}} = 1/e$$

- $t_{corr} = inverse gap.$
- $t_{\rm mix} \gg t_{\rm corr}$ leads to cutoff phenomenon.
- Aldous-Diaconis (1986)
- Diaconis-Fill-Pitman (1992)



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Detailed balance, global balance, lifting



• 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':

$$\pi_{a} = \sum_{b \in \Omega} \pi_{b} P_{ba}$$



Anharmonic potential (1D potential 1/6)



• Anharmonic potential

$$U_{24}(x) = \frac{x^2}{2} + \frac{x^4}{4} \tag{1}$$

Boltzmann distribution

$$\pi_{24}(x) = \exp\left[-\beta U_{24}(x)\right] = \pi_2(x)\pi_4(x),$$

with $\beta = 1/(kT)$: inverse temperature



Metropolis algorithm (1D potential 2/6)



• satisfies detailed balance ...

$$\pi_{24}(x)\min\left[1,\frac{\pi_{24}(x')}{\pi_{24}(x)}\right] = \pi_{24}(x')\min\left[1,\frac{\pi_{24}(x)}{\pi_{24}(x')}\right]$$

• ... famous, but not the end of history.



'Factorized' Metropolis (naive) (1D potential 3/6)



also satisfies detailed balance...

$$\pi_{24}(x)\min\left[1,\frac{\pi_{2}(x')}{\pi_{2}(x)}\right]\min\left[1,\frac{\pi_{4}(x')}{\pi_{4}(x)}\right] = \pi_{24}(x)\min\left[1,\frac{\pi_{2}(x')}{\pi_{2}(x)}\right]\min\left[1,\frac{\pi_{4}(x')}{\pi_{4}(x)}\right]$$

... but is 'naive'.

'Factorized' Metropolis (patched) (1D potential 4/6)



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$$\mathbb{P}\left\{\Upsilon_{2} < \min\left[1, \frac{\pi_{2}(x')}{\pi_{2}(x)}\right] \bigwedge \Upsilon_{4} < \min\left[1, \frac{\pi_{4}(x')}{\pi_{4}(x)}\right]\right\} = \\\min\left[1, \frac{\pi_{2}(x')}{\pi_{2}(x)}\right]\min\left[1, \frac{\pi_{4}(x')}{\pi_{4}(x)}\right]$$

• ... tautological statement, wide-ranging implications

'Lifted' Metropolis (1D potential 5/6)



• breaks detailed balance, satisfies global balance ...

• ... and implements a non-reversible Markov chain.

Peters, de With (2012), Bierkens, Fearnhead, Roberts (2019)



'Lifted factorized' Metropolis (1D potential 6/6)



- breaks detailed balance and is 'lifted'...
- ... coordinates independent factor decisions through consensus.



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Random walk (RW) on the one-dimensional lattice

• In the bulk:



• At the boundary:





Lifted random walk (I-RW)

• Lifting of samples:



(4)

upérieure

In the bulk:



• At the boundary:



Diaconis, Holmes, Neal (2000)

Random walk, lifted random walk (examples)

Symmetric simple exclusion process (SSEP)

• Move (first part ...)



• Move (... second part)



Totally asymmetric simple exclusion process (TASEP)



forward-backward coupling (ad-hoc, or boundary conditions).
 NB: Non-reversible, i.e. non-equilibrium, but samples equilibrium
 Boltzmann distribution.

Lifted TASEP (definition)

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



• Move (second part ...)



TASEP (example)

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

$$1 \equiv \overrightarrow{\bullet \bullet} \qquad 2 \equiv \overrightarrow{\bullet \bullet} \qquad 3 \equiv \overrightarrow{\bullet \bullet} \qquad 4 \equiv \overrightarrow{\bullet \bullet} \qquad 5 \equiv \overrightarrow{\bullet \bullet} \qquad 6 \equiv \overrightarrow{\bullet \bullet} = \overrightarrow{\bullet \bullet} \qquad 6 \equiv \overrightarrow{\bullet \bullet} = \overrightarrow{\bullet \bullet} \qquad 6 \equiv \overrightarrow{\bullet \bullet} = \overrightarrow{\bullet \bullet}$$



Lifted TASEP (example)

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



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Algorithm	mixing	relaxation (inverse gap)
SSEP	N ³ log N	N ³
TASEP	$N^{5/2}$	N ^{5/2}
Lifted TASEP (optimal $lpha$)	N^2	N ²

- continuous-space versions available (Kapfer-Krauth (2017))
- see Essler-Krauth (2023)



Parameter-dependence of relaxation times







- A second revolution in Markov-chain Monte Carlo underway.
- Time scales of MCMC much better understood.
- Non-reversible MCMC is what comes after the revolution.
- Exactly solvable non-trivial models of crucial importance.
- Lifted TASEP may serve as paradigm