Non-reversible sampling of hardcore particle systems

Manon Michel

CNRS, Laboratoire de mathématiques Blaise Pascal, Université Clermont-Auvergne

April 3rd-7th, 2023 Analysis and simulations of metastable systems Conference, CIRM, Marseille

Joint work with Arnaud Guillin, Tristan Guyon and Athina Monemvassitis (UCA)



Outline

Pitfalls of hardcore particle sampling

Non-reversible and continuous-time sampling by ECMC

Invariance through interplay of transport and direction changes

Ergodicity

Generalized deterministic flow

Anisotropic hardcore particles

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Equation of State Calculations by Fast Computing Machines

NICHOLAS METROPOLIS, ARIANNA W. ROSENBLUTH, MARSHALL N. ROSENBLUTH, AND AUGUSTA H. TELLER, Los Alamos Scientific Laboratory, Los Alamos, New Mexico

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



Diffusive dynamics

- ► Correlated sample: $\sigma^2(\bar{\Theta}) \propto \tau(\Theta)$ $C_{\Theta}(t) = \frac{\langle \Theta(t'+t)\Theta(t') \rangle - \langle \Theta^2 \rangle}{\langle \Theta^2 \rangle - \langle \Theta \rangle^2}$
- Around 2nd order phase transition $\tau \propto \xi^z \propto L^z$ $C_{\Theta}(t) \sim \exp(-t/\tau)$

Cluster algorithms (Swendsen and Wang (1987), Wolff (1989))











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Cluster algorithms (Swendsen-Wang et al (1987); Wolff et al (1989))

- ► Instead of single-spin flip, flip of correlated-spin clusters → reduction of critical slowing down
- Create or not a bond between neighboring spins.
- Pick a random value for each newly built cluster

Detailed balance is fulfilled: flipping is involutary $q(x',x)/q(x,x') = \pi(x')/\pi(x) =$ no rejection



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How to produce collective moves?

- Continuous state space. No discrete symmetry as for spin lattices to easily build global q.
- With detailed balance in hard-core particle systems: symmetric proposal probabilities are necessary for the scheme to be rejection-free.

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- Continuous state space. No discrete symmetry as for spin lattices to easily build global q.
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Break DB: Non-reversibility?



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Event-chain Monte Carlo

Bernard et al (2009)



Michel et al (2014), Kapfer et al (2015)



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Event-chain Monte Carlo



- ► (a) A sphere is updated ballistically.
- (b-c) At some point, an *event* with another sphere is triggered and the ballistic flow is updated.

$$a_{\mathsf{Met}} = \exp(-[\sum_j \delta E_{ij}]_+) \to a_{\mathsf{Fact}} = \exp(-\sum_j [\delta E_{ij}]_+) \xrightarrow[\delta \to 0]{} 1 - \beta \sum_j [\mathrm{d} E_{ij}]_+$$

• (d-e) A *refreshment* is triggered and the ballistic flow is updated.

ECMC is rejection-free and relies on a control by the events of the ballistic exploration to ensure the correct invariant distribution.

Piecewise Deterministic Markov Process



PDMP characterizing elements (Davis (1993), in MCMC: Bouchard-Côté et al (2018), Bierkens et al (2018))

- ▶ Differential flow $(\phi_t)_{t \ge 0}$
- Jump rate $\lambda(x, e) + \overline{\lambda}$
- Markov kernel Q

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Michel et al (2014), Monemvassitis et al (2023)



▶ Particle systems,
$$E(x) = \sum_{i,j} E_{ij}(x)$$
,
 $v = (e, i) \in \{(0, 1), (1, 0)\} \times \{1...N\}$

▶ Differential flow $\phi_t(x, (e, i)) = (\{x_1, ..., \mathbf{x_i} + \mathbf{te}, ..., x_N\}, (v, i))$



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- Markov kernel $Q_j((x, (e, i)), di'de') = \delta(j i')\delta(e e')de'di'$



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Soft spheres

$$\mathcal{A}f(x,(e,i)) = \langle \nabla_{x_i}f(x,(e,i)) + \sum_{k=1}^N \lambda_k(x,(e,i)) \left(\int_{\mathcal{V}} Q_k((x,(e,i)), \mathrm{d}e', \mathrm{d}i')f(x,(e',i')) - f(x,(e,i)) \right)$$

Michel et al (2014), Monemvassitis et al (2023)



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Hard spheres = boundary effect

$$\begin{aligned} \mathcal{A}f(x,(e,i)) &= \langle \nabla_{x_i}f(x,(e,i)), e \rangle, e \rangle, (x,(e,i) \in E \\ Q^b((x,(e,i)), A) &= \sum_{k=1, \neq i}^N \int_{\mathcal{V}} \mathbb{1}_A(x,(e',i'))Q_k(x,(e,i), \mathrm{d}e', \mathrm{d}i'), (x,(e,i)) \in \partial E \end{aligned}$$

Michel et al (2014), Monemvassitis et al (2023)

Common: exponential refreshment $(x, v) \in E = \Omega \times \mathcal{V}$

$$\begin{cases} \mathcal{A}f(x,v) = & \langle \nabla_x f(x,v),v \rangle + \lambda(x,v) \left(\int_{\mathcal{V}} \mathcal{Q}((x,v),\mathrm{d}v')f(x,v') - f(x,v) \right) \\ & +\lambda_r \left(\int_{\mathcal{V}} f(x,v')\mathrm{d}\mu_{\mathcal{V}}(v') - f(x,v) \right), (x,v) \in E \end{cases} \\ \mathcal{Q}^b((x,v),\mathcal{A}) = & \int_{\mathcal{V}} \mathbb{1}_{\mathcal{A}}(x,v')\mathcal{K}(v,\mathrm{d}v'), (x,v) \in \partial E \end{cases}$$

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 \rightarrow Describe it as a boundary effect, $l \in \mathcal{L}$, $\partial \mathcal{L} = 0$, $(x, v, l) \in E = \Omega \times \mathcal{V} \times \mathcal{L}$

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$$\begin{aligned} \mathcal{A}_{\mathsf{Ref}}f(x,v,l) &= \langle \nabla_{x}f(x,v,l),v \rangle - \partial_{l}f(x,v,l) \\ &+ \lambda(x,v) \left(\int_{\mathcal{V}} Q((x,v), \mathrm{d}v')f(x,v',l) - f(x,v,l) \right), (x,v,l) \in E \\ Q^{b}_{\mathsf{Ref}}((x,l,v),A) &= \quad \mathbb{1}_{\partial \mathcal{L}} \int_{\mathcal{L}} \mathbb{1}_{A}(x,l',v')R(l,\mathrm{d}l')\mathrm{d}\mu_{\mathcal{V}}(v') \\ &+ (1 - \mathbb{1}_{\partial \mathcal{L}}) \int_{\mathcal{V}} \mathbb{1}_{A}(x,l,v')K(v,\mathrm{d}v'), (x,v,l) \in \partial E \end{aligned}$$

Monemvassitis et al (2023)

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$$\begin{split} \mathcal{A}_{\mathsf{Ref}}f(x,v,l) &= \langle \nabla_{x}f(x,v,l),v\rangle - \partial_{l}f(x,v,l) \\ &+ \lambda(x,v) \left(\int_{\mathcal{V}} \mathcal{Q}((x,v),\mathrm{d}v')f(x,v',l) - f(x,v,l) \right), (x,v,l) \in E \\ \mathcal{Q}_{\mathsf{Ref}}^{b}((x,l,v),A) &= \quad \mathbb{1}_{\partial\mathcal{L}} \int_{\mathcal{L}} \mathbb{1}_{A}(x,l',v') \mathcal{R}(l,\mathrm{d}l')\mathrm{d}\mu_{\mathcal{V}}(v') \\ &+ (1 - \mathbb{1}_{\partial\mathcal{L}}) \int_{\mathcal{V}} \mathbb{1}_{A}(x,l,v') \mathcal{K}(v,\mathrm{d}v'), (x,v,l) \in \partial E \end{split}$$

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Conditions on R and μ_L : Transport by $-\partial_I$ compensated by R-jump

$$\mu_L(0^+)R(0,\mathrm{d} I) = (-\partial_I \mu_L(I))\mathrm{d} I \iff \mu_L(I) = h(I)\mathbb{1}_L,$$

where $L \subset \mathcal{L}$ and h decreasing over L so that $h(0^+) > 0$, $\int_{\mathcal{L}} \frac{-\partial_l \mu_L(l)}{\mu_L(0^+)} dl = 1$

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Recovering exponential and fixed-time refreshment

 $\begin{cases} \text{ fixed-time } T: \ \mu_L(l) = \frac{1}{T} \mathbb{1}_{]0, \ T[} \text{and } R(l, \mathrm{d}l') = \delta(l' - T) \mathrm{d}l', \text{ with } T > 0, \ L =]0, \ T] \\ \text{ exponential of rate } \lambda_r: \ \mu_L(l) = \lambda_r e^{-\lambda_r l} \mathbb{1}_{l>0} \text{and } R(l, \mathrm{d}l') = \lambda_r e^{-\lambda_r l'} \mathbb{1}_{l'>0} \mathrm{d}l', \text{ with } \lambda_r > 0, \ L = \mathcal{L} \end{cases}$

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▶
$$\mu_L(I) = A(T-I)^k \mathbb{1}_{0 < I \le T}$$
, and refreshment $R(0, dI) = \frac{k}{T^k} (T-I)^{k-1} \mathbb{1}_{0 < I \le T}$,
▶ $\mu_L(I) = \frac{A}{(T+I)^k} \mathbb{1}_{0 < I}$, and refreshment $R(0, dI) = \frac{kT^k}{(T+I)^{k+1}} \mathbb{1}_{0 < I}$,
▶ $\mu_L(I) = Ae^{-TI^k} \mathbb{1}_{0 < I}$, and refreshment $R(0, dI) = kTI^{k-1}e^{-TI^k} \mathbb{1}_{0 < I}$, etc
where $(T, k) \in \mathbb{R}^{*2}_+$, $I \in \mathcal{L}$ and A is a suitable normalization constant. Monemvassitis et al (2023)

Invariance: Transport compensated by the direction changes

Infinitesimal generator
$$\mathcal{A}f = \lim_{t \to 0} \frac{P_t f - f}{t}$$

 $\mathcal{A}f = \underbrace{D_{\phi}f(x, e)}_{\text{Transport}} + \underbrace{\lambda(x, e) \int_{\mathcal{V}} (f(x, e') - f(x, e))Q((x, e), de')}_{\text{Events - Direction changes}} + \underbrace{\overline{\lambda} \int_{\mathcal{V}} (f(x, e') - f(x, e))\mu(de')}_{\text{Refreshment}}$
Conditions for $\tilde{\pi} = \pi \times \mu$ invariant: $\int_{\Omega \times \mathcal{V}} \mathcal{A}f d\pi d\mu = 0$
 $\int_{\Omega \times \mathcal{V}} D_{\phi}f(x, e)\pi(dx)\mu(de)$
 $= \int_{\Omega \times \mathcal{V}} \int_{\mathcal{V}} \lambda(x, e)(f(x, e') - f(x, e))Q((x, e), de')\pi(dx)\mu(de)$

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 $= \int_{\Omega \times \mathcal{V}} \int_{\mathcal{V}} \lambda(x, e)(f(x, e') - f(x, e))Q((x, e), de')\pi(dx)\mu(de)$
With a flow along e, by integration by part, $(\pi(x) \propto \exp(-E(x))$
 $\int_{\mathcal{V}} \langle \nabla E(x), -e' \rangle_{+} f(x, e')\mu(de') = \int_{\mathcal{V}} \int_{\mathcal{V}} \langle \nabla E(x), e \rangle_{+} f(x, e')Q((x, e), de')\mu(de)$

16/30

Designing events through global symmetry exploitation

$$\underbrace{\int_{\mathcal{V}} \langle \nabla E(x), -e' \rangle_{+} f(x, e') \mu(\mathrm{d}e')}_{\text{brought by transport}} = \underbrace{\int_{\mathcal{V}} \int_{\mathcal{V}} \langle \nabla E(x), e' \rangle_{+} f(x, e) Q((x, e'), \mathrm{d}e) \mu(\mathrm{d}e')}_{\text{redistribution by direction change}}$$

Exploitation of mirror symmetry through factorization (Michel et al 2014)

$$\boldsymbol{\nabla}_{x_i} E_{ij}(x) = -\boldsymbol{\nabla}_{x_j} E_{ij}(x) \text{ (i.e. } \operatorname{div} E_{ij} = 0) \rightarrow Q_j((x, (e, i)), \operatorname{di'}\operatorname{d} e') = \delta(j - i')\delta(e - e')\operatorname{d} e'\operatorname{d} i' \\ \rightarrow Q_j((x, (e, i)), \operatorname{di'}\operatorname{d} e') = \delta(j - i')\delta(e - R_x(e'))\operatorname{d} e'\operatorname{d} i'$$

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Exploitation of translational invariance (Harland et al 2017)

$$\nabla \cdot E = 0 \rightarrow Q((x, (e, i)), \mathrm{d}i' \mathrm{d}e') = \sum_{k} \frac{\langle \nabla_{x_k} E, e \rangle_-}{\sum_j \langle \nabla_{x_j} E, e \rangle_+} \delta(e - e') \delta(k - i') \mathrm{d}e' \mathrm{d}i'$$

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Exploitation of rotational invariance (Michel et al 2020)

~

$$\int \langle \nabla E, e \rangle \mu(\mathrm{d} e) = 0 \to Q((x, (e, i)), \mathrm{d} i' \mathrm{d} e') \propto \langle \nabla E, e' \rangle_{-} \mu(\mathrm{d} e')(\ldots \mathrm{d} i')$$

Outline

Pitfalls of hardcore particle sampling

Non-reversible and continuous-time sampling by ECMC

Invariance through interplay of transport and direction changes

Ergodicity

Generalized deterministic flow

Anisotropic hardcore particles

Ergodicity in sphere systems

Ergodicity in PDMP

Goal is to find a density of paths connecting two states, while probability minorization (positive Harris recurrent, irreducible skeleton chain, Meyn and Tweedie (1993)). Gain randomness through the jump or refreshment times.

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In soft/hard-sphere systems

- \blacktriangleright Dealing with diverging $\lambda \rightarrow$ control of a minimal distance
- ▶ Dealing with hardcore conditions \rightarrow density condition (\sim can pack 3N spheres of radius d_{pair}) (Metropolis algorithm, Diaconis, Lebeau, Michel (2011): linear)
- Dealing with periodicity and $e = +u_x, +u_y$

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Strategy $x_0 \rightarrow x_f$

Start in the case of well-separated spheres
 (min_{i≠j}(d(x_{0,i}, x_{0,j}), d(x_{0,i}, x_{f,j}), d(x_{f,i}, x_{f,j})) > 2d_{pair}) (d=periodic distance).
 Show the possibility to define a valid path only along +u_x and +u_y from any path along
 ±u_x, u_y which directly connects x₀ and x_f.

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Can ease the dependence on x_0 (only travel times) for soft spheres.

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Upgrading the deterministic flow

General flow: differential drift ϕ , $\phi_{t+s} = \phi_t \circ \phi_s$

$$rac{\mathrm{d}(x_t,v_t)}{\mathrm{d}t} = \phi(x_t,v_t)$$





 $\int \mathcal{A}f(x,\nu)\mathrm{d}\pi(x)\mathrm{d}\mu(\nu) \to \int \mathrm{d}\pi(x)\mathrm{d}\mu(\nu)f(x,\nu)\nabla\cdot\phi(x,\nu) + \text{transport} + \text{events}$

Vanetti et al (2017)

Rotational flow, ϕ_R : α - rotation of *i*-th sphere around a point at $x_i - l(\cos(\psi), \sin(\psi))$ $(x, v) = (x, \alpha, l, \psi)$ and $\phi(x, v) = (l(-\sin\psi, \cos\psi)), 0, 0, \alpha) \rightarrow \nabla \cdot \phi_R(x, v) = 0$

Hybrid flow: alternate between ϕ_R and ϕ_T depending on ω (x, v) = ($x, \omega, (e, i), \alpha, l, \psi$) and $\phi_H(x, v) = \omega(e, 0, 0, 0, 0) + (1 - \omega)(l(-\sin\psi, \cos\psi)), 0, 0, 0, \alpha) \rightarrow \nabla \cdot \phi_H(x, v) = 0$

Guyon et al (2023)

Applications to hard disks - at hexatic density



Observable: ϕ_6 , N = 256

Guyon et al (2023)

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Non-reversible sampling of anisotropic particles

Without rotational flow



Elastic length



Tethered interaction



Höllmer et al (2021)

Rotations are necessary to thermalize dimers





Self-Rotations cannot be naively propagated as translations, as breaking of symmetry: $|d_{R(i)}E_{ij}| \neq |d_{R(j)}E_{ij}|$ in general





No backtracking!



Numerical comparaison - (density $\rho = 0.7$, N = 32)



Polarisation vector

Nematic order parameter

Guyon et al (2023)

Numerical comparaison - (density $\rho = 0.5$, N = 32)



Polarisation vector

Nematic vector order parameter

Guyon et al (2023)

Conclusion

Finding the entropic opening by building persistency into moves along PDMP

- Non-reversibility obtained by exploiting global symmetries
 - Flexible schemes based on the exploitation and (stochastic) control of a ballistic exploration
 of the state space.
 - The PDMP framework allows for a clear and direct formalism.
 - Generalisation to other flows than the translations, generalisation to the non-reversible sampling of anisotropic particle systems.

Some questions

- Ergodicity proof at relevant densities, with/without refreshment
- Trade off between generating persistent transport while avoiding building too strong correlations
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Thank you for your attention!



Joint work with Arnaud Guillin, Tristan Guyon, Athina Monemvassitis (UCA)

References

- M. Benaïm, S. Le Borgne, F. Malrieu, and P.-A. Zitt. Annales de l'Institut Henri Poincaré, Probabilités et Statistiques, 51(3), 2015.
- E. P. Bernard, W. Krauth, D. B. Wilson Phys. Rev. E 80 056704 (2009)
- ▶ J. Bierkens, G. O Roberts, P.-A. Zitt, et al. *The Annals of Applied Probability*, **29**(4):2266–2301, 2019.
- J. Bierkens, P. Fearnhead, G. Roberts, Ann. Statist., 47(3), 1288-1320 (2019)
- A. Bouchard-Côté, S. Vollmer and A. Doucet. JASA, 113(522): 855-867 (2018)
- M. Davis, Markov Models & Optimization, Volume 49. CRC Press. (1993)
- P. Diaconis, G. Lebeau, and L. Michel. Inventiones Mathematicae, 185(2):239–281, 2011.
- J. Harland, M. Michel, T. A. Kampmann and J. Kierfeld, Phys. Rev. E, 117 (3), 30001 (2017).
- M. Klement, S. Lee, J. A. Anderson, and M. Engel, *Journal of Chemical Theory and Computation* 17, 4686 (2021).
- M. Michel, S. C. Kapfer and W. Krauth, J. Chem. Phys., 140, 054116 (2014)
- M. Michel, A. Durmus and S. Sénécal, JCGS, 29(4): 689–702 (2020).
- A. Monemvassitis, A. Guillin, M. Michel, Journal of Statistical Physics 190, 66 (2023).
- P. Vanetti, A. Bouchard-Côté, G. Deligiannidis, A. Doucet, arxiv 1707.05296