# Phase transitions, metastability and critical slowing down in statistical physics 

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Both research fields estimate expectations wrt some probability distribution $\pi(x ; \beta, \theta) \propto e^{-\beta U(x ; \theta)}$

## Bayesian inference

- $\pi(x \mid y, \beta, \theta) \propto e^{-\beta U(x \mid y, \theta)}$
- Fix hyperparameters $\boldsymbol{\beta}, \boldsymbol{\theta}$.
- Encode input data $y$ via likelihood...
- ...and estimate expectations wrt $\boldsymbol{x}$.


## Statistical physics

- $\pi(x ; \beta, \theta) \propto e^{-\beta U(x ; \theta)}$
- Defined independent of input data.
- Expectations are functions of $\boldsymbol{\beta}$ and $\boldsymbol{\theta}$.
- $\boldsymbol{\beta}, \boldsymbol{\theta}$ are thermodynamic parameters, eg, system temperature $\equiv 1 / \beta$.



## P!|OS Oł P!nb!

## Expected potential variance per particle

 (black) is non-analytic at $\boldsymbol{\beta}_{\mathrm{c}}$
magnetic to non-magnetic


# - Statistical physics and phase transitions 

- Metastability and Wolff algorithm
- Continuous state spaces and ECMC




## Periodic boundary conditions



## Thermodynamic phase space

- With $\chi(x ; \beta, \theta, N)$ some observable...
- Thermodynamic phase space (TPS) of $\chi(x ; \beta, \theta, N)$ is $\lim _{N \rightarrow \infty} \mathbb{E}[\chi(x ; \beta, \theta, N)]$ as a function of $\boldsymbol{\beta}$ and $\boldsymbol{\theta}$.
- Thermodynamic phase: any open and connected region of TPS where $\lim _{N \rightarrow \infty} \mathbb{E}[\chi(\boldsymbol{x} ; \boldsymbol{\beta}, \boldsymbol{\theta}, N)]$ is analytic.
- Phase transition: any boundary between any two thermodynamic phases.

2D Ising model has non-analytic expectations at $\beta=\beta_{c} \ldots$

...but no phase transition has been detected in 1D case


$$
U_{\text {Ising }}=-\frac{J}{2} \sum_{i=1}^{N} \sum_{j \in S_{i}} x_{i} x_{j}, x_{i}= \pm 1
$$

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- Single active particle $a \in\{1, \ldots, N\}$
- $x_{a}^{\prime}=-x_{a}$
- $\Delta U_{\text {Ising }}=J \sum_{j \in S_{a}} x_{a} x_{j}$
- Accept $x_{a}^{\prime} \mathbf{w} /$ prob $\min \left[1, \exp \left(-\beta \Delta U_{\text {Ising }}\right)\right]$
- NB, one unit of MC time corresponds to $N$ attempted particle moves
- Magnetisation: $m(x ; \beta, J, h, N):=\frac{1}{N} \sum_{i=1}^{N} x_{i}$
- $\mathbb{E}[\boldsymbol{m}(\boldsymbol{x} ; \boldsymbol{\beta}, \boldsymbol{h}=\mathbf{0}, N)]=\mathbf{0}$ for all $\boldsymbol{\beta}<\infty$ (spin-flip symmetry)
- So $\frac{1}{\tau_{n}} \sum_{t=\tau_{1}}^{\tau_{n}} m\left(x_{t} ; \beta J, h=0, N\right) \rightarrow \mathbf{0}$ on some timescale $\tau_{n}$
- But at low temperature and w/Metropolis dynamics...
- ... $\tau_{n}$ diverges with system size $N$


## Low-temp Metrop dynamics freeze...


...as neighbours are typically aligned

## Wolff algorithm



## Flips entire clusters of aligned spins in 'intelligent' way

1. Randomly pick base lattice site for new cluster
2. Add aligned neighbours (to cluster) with probability $p:=1-e^{-2 \beta J}$
3. Repeat step 2 for each new spin...
4. ...and flip entire cluster with probability one.


Magnetisation: $\boldsymbol{m}(\boldsymbol{x} ; \boldsymbol{\beta}, \boldsymbol{J}, \boldsymbol{h}, N):=\frac{1}{N} \sum_{i=1}^{N} x_{i}$


## Experimental-theoretical discrepancies are essence of symmetry breaking



## Thermodynamic limit of expected magnetisation

is singular for all $\beta>\beta_{\text {Ising }}$

$\lim _{h \downarrow 0} \lim _{N \rightarrow \infty} \mathbb{E} \boldsymbol{m}(\boldsymbol{x} ; \boldsymbol{\beta J}, \boldsymbol{h}, N) \neq 0=\lim _{N \rightarrow \infty} \lim _{h \downarrow 0} \mathbb{E} \boldsymbol{m}(\boldsymbol{x} ; \boldsymbol{\beta J}, \boldsymbol{h}, N)$






High-temperature unimodal


Fig. (c) is now near the transition where growing correlated clusters induce strong autocorrelations






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## ...were first applied to hard disks





## Event-chain Monte Carlo

## Chain of events traverses system



## Random coordinate switches...



## ...to sample along both dimensions



## ECMC solved 2D melting



Disputed for 50 years

## Colloidal hard-disk experiment



Confirms ECMC numerics

## Continuous potentials ${ }^{1}$

- $\pi_{\text {hard }}(x)=0$ or const. $\Leftrightarrow U_{\text {hard }}(x)=\infty$ or finite...
- ...so ECMC freely advances hard disks until $\mathrm{dU} / \mathrm{d} x_{a}=\infty$...
- ...but particles never collide in the case of continuous potentials $U(x)$
- $\Rightarrow$ somehow account for continuous increases in $U(x)$ ?
- Consider $m$ Metropolis translations of length $\Delta$ in a fixed direction.
- Probability of translating active particle $a$ through distance $\eta:=m \Delta$ is...
- $p\left(x_{a} \rightarrow x_{a}+\eta\right)=\prod_{i=1}^{m} \min \left[1, \exp \left(-\beta\left[U\left(x_{a}+\Delta i\right)-U\left(x_{a}+\Delta(i-1)\right)\right]\right)\right]$

$$
\begin{aligned}
& =\exp \left[-\beta \sum_{i=1}^{m} \max \left(0, U\left(x_{a}+\Delta i\right)-U\left(x_{a}+\Delta(i-1)\right)\right)\right] \\
& \rightarrow \exp \left[-\beta \int_{0}^{\eta} \max \left(0, \nabla_{a} U(x)\right) \mathrm{d} x_{a}\right] \text { as } \Delta \rightarrow 0
\end{aligned}
$$

- $\Rightarrow$ Advance active particle at constant velocity $v$ from time $t_{0} \geq 0$ and solve:

$$
-\log \Upsilon=\beta \int_{t_{0}}^{t_{\eta}} \max \left(0, v \cdot \nabla_{a} U(x)\right) \text { dt where } \Upsilon \sim \mathcal{U}[0,1) \ldots
$$

- ...to find the next event time $\boldsymbol{t}_{\boldsymbol{\eta}}:=\boldsymbol{t}_{\mathbf{0}}+\boldsymbol{\eta} / \boldsymbol{v}$ (assuming no 'boundary' collisions).
- Particle $i$ then becomes active $w /$ prob. $\propto \max \left(0,-v \cdot \nabla_{i} U\left[x\left(t_{\eta}\right)\right]\right)$ at $t=t_{\eta}$.
- Need to integrate $v \cdot \nabla_{a} U(x)$ over only positive contributions...
- ...but this is non-trivial for multiple particles
- So we have two options for Poisson process (PP):

1. Thinned PP: choose $\widetilde{q_{a}}$ to overestimate event rate $q_{a}(x):=\beta \max \left(0, v \cdot \nabla_{a} U(x)\right)$, then confirm events with probability $q_{a}(x) / \widetilde{q_{a}}(x)$
2. 2-particle blocking: Sample Poisson process of each two-particle interaction and take shortest displacement (superposition of PPs)



$$
U_{\mathrm{xy}}=-J \sum_{i=1}^{N} \sum_{j \in S_{i}} \cos \left(x_{i}-x_{j}\right) \text { with } x_{i} \in(-\pi, \pi], J>0
$$



$$
U_{\mathrm{XY}}=-J \sum_{i=1}^{N} \sum_{j \in S_{i}} \cos \left(x_{i}-x_{j}\right) \text { with } x_{i} \in(-\pi, \pi], J>0
$$



$$
m(x ; \beta, J, h, N):=\frac{1}{N} \sum_{i=1}^{N}\left(\cos x_{i}, \sin x_{i}\right)^{t}, x_{i} \in(-\pi, \pi]
$$




ECMC's constant-speed dynamics circumvent critical slowing down?

## Summary and outlook

- Bayesians fix hyperparameters, whereas physicists vary them.
- Varying hyperparameters can induce metastability and critical slowing down.
- Physicists combat these phenomena w/sophisticated sampling algorithms.
- Future plans: use ECMC to characterise CSD in 2DXY model; explore Bayesian analogues.
- Also interested in $\pi$-invariance of canonical ECMC if anyone has any ideas!
- Thanks to Sam Livingstone ${ }^{1}$, EPSRC and Advanced Computing Research Centre (Bristol).

- $\boldsymbol{U}_{\text {lsing }}=-\frac{J}{2} \sum_{i=1}^{N} \sum_{j \in S_{i}} x_{i} x_{j}, x_{i}= \pm 1$
- Spin-spin correlation length increases as temperature decreases
- $\rightarrow$ nonergodic Metropolis dynamics
- Wolff combats this by flipping clusters


## Fundamental axiom

- If some scalar observable $\chi(x ; \beta, \theta, N)$ is sum of $O(N)$ random numbers...
- ...and $\frac{\sigma_{\chi}}{\mathbb{E}[\chi]}$ can be made arbitrarily small as $N \rightarrow \infty\left(\right.$ with $\left.\lim _{N \rightarrow \infty} \mathbb{E}[\chi(x ; \beta, \theta, N)] \neq 0\right)$...
- ...then $\exists N_{0} \in \mathbb{N}$ s.t. $\left|\frac{\mathbb{E}\left[\chi\left(x ; \beta, \theta, N=N_{0}\right)\right]}{\lim _{N \rightarrow \infty} \mathbb{E}[\chi(x ; \beta, \theta, N)]}-1\right|<\varepsilon$ (with $\varepsilon>0$ immeasurably small)
- $\Rightarrow$ thermodynamic limit (usually!) reflects macroscopic physics



## No phase transition wrt free energy, F

- $U_{\text {Ising }}=-\frac{J}{2} \sum_{i=1}^{N} \sum_{j \in S_{i}} x_{i} x_{j}-h \sum_{i=1}^{N} x_{i}, h \in \mathbb{R}$
- $F_{\text {Ising }}^{d=1}(\beta, J, h, N)=-\beta^{-1} \log \left[\lambda_{+}^{N}(\beta, J, h)+\lambda_{-}^{N}(\beta, J, h)\right]$
- $\lambda_{ \pm}^{N}(\beta, J, h):=e^{\beta J}\left[\cosh (\beta h) \pm \sqrt{\sinh ^{2}(\beta h)+e^{-4 \beta J}}\right]$


## 2D Ising model

- Expected specific heat $\left(\mathbb{E} C_{V}=\beta^{2} \operatorname{Var}[U]\right)$ is nonanalytic as $\boldsymbol{N} \rightarrow \infty$ at $\boldsymbol{\beta}=\boldsymbol{\beta}_{\mathrm{c}}, \boldsymbol{h}=\mathbf{0}$ (black curve)
- $\lim _{N \rightarrow \infty} \frac{\mathbb{E} C_{V}(x ; \beta, J, h=0, N)}{N}=\beta^{2} \partial_{\beta}^{2} \gamma(\beta J)$
- $\gamma(\beta):=\ln [2 \cosh (2 \beta J)]+\frac{1}{\pi} \int_{0}^{\pi / 2} \ln \left[\frac{1}{2}\left(1+\sqrt{1-\frac{4 \sinh ^{2}(2 \beta) / \sin ^{2}(w)}{\cosh ^{4}(2 \beta)}}\right)\right] \mathrm{d} w$

Thermodynamic specific heat per particle (black curve) diverges at $\boldsymbol{\beta}=\boldsymbol{\beta}_{\mathrm{c}}, \boldsymbol{h}=\mathbf{0}$


## What about order and magnetisation?

- $m(x ; \beta, J, h, N):=\frac{1}{N} \sum_{i=1}^{N} x_{i}$
- $\boldsymbol{m}_{\mathbf{0}}(\boldsymbol{\beta}, \boldsymbol{J}):=\lim _{h \downarrow \mathbf{0}} \lim _{N \rightarrow \infty} \mathbb{E} \boldsymbol{m}(\boldsymbol{x} ; \boldsymbol{\beta}, \boldsymbol{J}, \boldsymbol{h}, N)$ is...
- ...also non-analytic at $\beta=\beta_{\text {c }}$ (below $\&$ red curve)
- $m_{0}(\beta, J)=\left\{\begin{array}{lr}\left(1-(\sinh (2 \beta J))^{-4}\right)^{1 / 8} \text { for } \beta>\boldsymbol{\beta}_{\mathrm{c}} \\ 0 & \text { for } \beta<\boldsymbol{\beta}_{\mathrm{c}}\end{array}\right.$

Spontaneous magnetisation ( $m_{0}$ in red) is also non-analytic...

...indicating an order-disorder transition at $\boldsymbol{\beta}=\boldsymbol{\beta}_{\mathrm{c}}$



$$
\boldsymbol{m}_{\mathbf{0}}(\boldsymbol{\beta}, \boldsymbol{J}):=\lim _{h \downarrow \mathbf{0} N \rightarrow \infty} \lim _{N} \mathbb{E} \boldsymbol{m}(\boldsymbol{x} ; \boldsymbol{\beta}, \boldsymbol{J}, \boldsymbol{h}, N)
$$



Thermodynamic limit is singular as swapping limits in equation returns zero


$$
\boldsymbol{m}_{\mathbf{0}}(\boldsymbol{\beta}, \boldsymbol{J}):=\lim _{h \downarrow 0} \lim _{N \rightarrow \infty} \mathbb{E} \boldsymbol{m}(\boldsymbol{x} ; \boldsymbol{\beta}, \boldsymbol{J}, \boldsymbol{h}, \boldsymbol{N})
$$

## Translational symmetry

- ECMC potentials: symmetric to simultaneous translation of both particles;
- $\mathrm{U}\left(\mathrm{x}_{\mathrm{i}}, x_{\mathrm{j}}\right)=\mathrm{f}(x)$;
- $x:=\left(x_{i}-x_{j}+L / 2\right) \bmod (L)-L / 2$ is shortest separation with PBCs.

1D, two-particle model

Time-driven, reversible algorithm


Red / blue: +ve / -ve x evolution

## Lifted Markov process

- Can explore $x$ via positive particle motion;
- Active particles augment the configuration space: $x \rightarrow(x, \xi= \pm 1)$;
- Lifting variable $\xi= \pm 1$ describes two copies of the original config. space (x);
- $\pi(x, \xi=1)=\frac{1}{2} \pi(x)=\pi(x, \xi=-1)$;
- Red: particle i active $\Rightarrow \xi=+1$; system on positive copy of config. space;
- Blue: particle j active $>\xi=-1$; system on


## Two copies of space



Red / blue: +ve / -ve x evolution negative copy of config. space;


## Nonreversible process


$\pi$-invariant?


At $(x, \xi= \pm 1): \mathrm{p}(x \rightarrow x+\xi)=\min \{1, \pi(x+\xi, \xi) / \pi(x, \xi)\}$



## Metropolis is very successful

- Easy to implement.
- Converges quickly enough in many settings.
- Recreates physical Brownian dynamics - useful for experiment.


## However...

- Convergence slow at high particle density with long-range interactions.
- Suffers from symmetry breaking.
- And critical slowing down - inducing strongly non-convergent estimates.

- Molecular dynamics (MD) follows numerical Newtonian trajectories (eg, red curve on potential landscape)
- It sets random initial particle positions and velocities...
- ...then solves $\ddot{x}_{i}=-\nabla_{i} U(x) \forall i$ at each time step.
- Approximately converges on $\pi$ w/resampled velocities.

- MD is typically much more efficient than Metropolis...
- ...and captures physical Newtonian dynamics.
- BUT it's unstable - especially at high particle density with long-range interactions...
- ...and it also suffers from energy drifts.

