

MATH327: Statistical Physics

Friday, 10 May 2024

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Something to consider

Suppose we want to *approximately* compute expectation values

$$\langle O \rangle = \sum_{i=1}^M O_i p_i = \frac{1}{Z} \sum_{i=1}^M O_i e^{-\beta E_i} = \frac{\sum_{i=1}^M O_i e^{-\beta E_i}}{\sum_{i=1}^M e^{-\beta E_i}}.$$

How many of the system's micro-states do we really need to analyze?

Plan

Mean-field approx. for Ising model

Numerical methods (more broadly applicable)

Recap

~~Assume~~ Assume on average small fluctuations around mean spin $\langle m \rangle$

→ non-interacting system w/ effective magnetic field H_{eff}

$$E = - \sum_{\langle ij \rangle} J_{ij} S_i S_j - H \sum_n S_n \rightarrow d \cdot N \langle m \rangle^2 - (Zd \langle m \rangle + H) \sum_n S_n$$

Self-consistency condition $\langle m \rangle = \tanh(\beta(Zd \langle m \rangle + H))$

$H=0$ & low $T \rightarrow$ three solutions $\langle m \rangle = 0, \pm m_0$

which is true?

Imagine perturbing $\langle m \rangle = 0 \rightarrow \langle m \rangle = \epsilon > 0$

$T=8$: $\langle m \rangle$ too large compared to $\tanh$

$\rightarrow$ decrease to stable $\langle m \rangle = 0$

$T=2$: $\langle m \rangle$ too small compared to $\tanh$

$\rightarrow$ increase to stable $\langle m \rangle = m_0 > 0$

Similarly go to $-m_0 \neq 0$ from $-\epsilon < 0$

$\therefore \langle m \rangle = 0$ unstable $\rightarrow \langle m \rangle \neq 0$ ordered phase ✓

Another way to see (in)stability

Plot $\tanh(2\beta d \langle m \rangle) - \langle m \rangle$ to find zeros

Positive $\rightarrow \langle m \rangle$ too small

Negative $\rightarrow \langle m \rangle$ too large

So mean-field approx $\rightarrow$ ordered/disordered like Full Ising model ✓

When does $\langle m \rangle = 0$ become unstable?

Need $\tanh > \langle m \rangle$ for $\langle m \rangle = \epsilon > 0 \rightarrow \text{slope} > 1$ at $\langle m \rangle = 0$

$$\left. \frac{d}{d\langle m \rangle} \tanh(2\beta d \langle m \rangle) \right|_{\langle m \rangle = 0} = \frac{d}{d\langle m \rangle} [2\beta d \langle m \rangle + \mathcal{O}(\langle m \rangle^3)]_{\langle m \rangle = 0}$$
$$= 2\beta d = 1 \rightarrow T_c = \frac{1}{\beta_c} = 2d$$

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Is this a phase transition (w/discontinuity)?

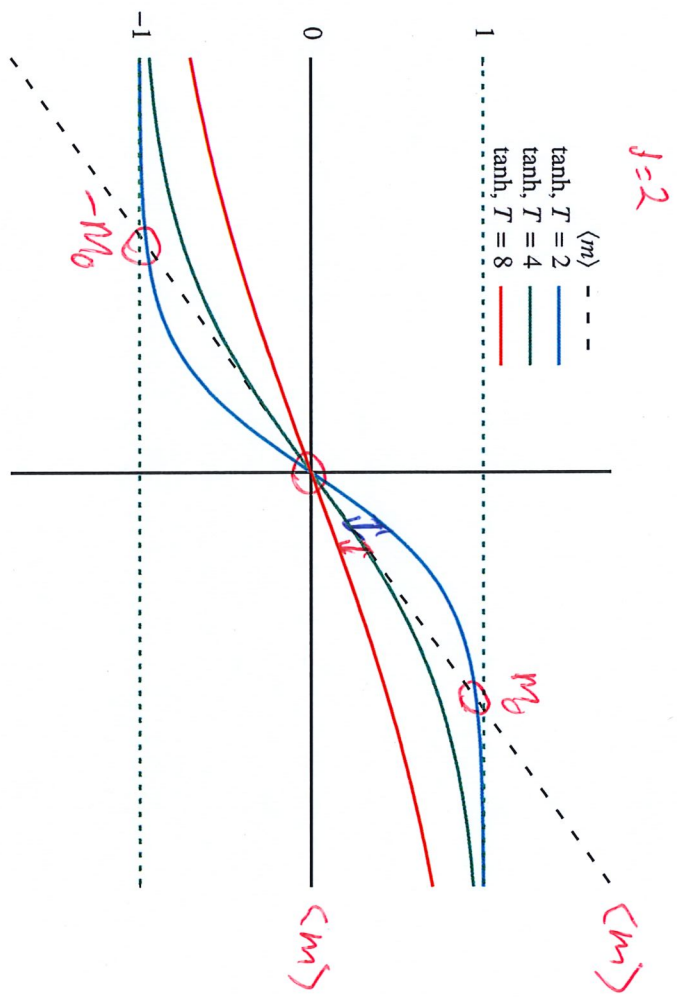
Consider $T \lesssim T_c \rightarrow 0 < |\langle m \rangle| \ll 1$

$$\langle m \rangle = \tanh(2\beta d \langle m \rangle) = 2\beta d \langle m \rangle - \frac{1}{3} (2\beta d \langle m \rangle)^3 + \mathcal{O}(\langle m \rangle^5)$$

$$\frac{1}{3} \left(\frac{T_c}{T} \right)^3 \langle m \rangle^2 = \frac{T_c}{T} - 1$$

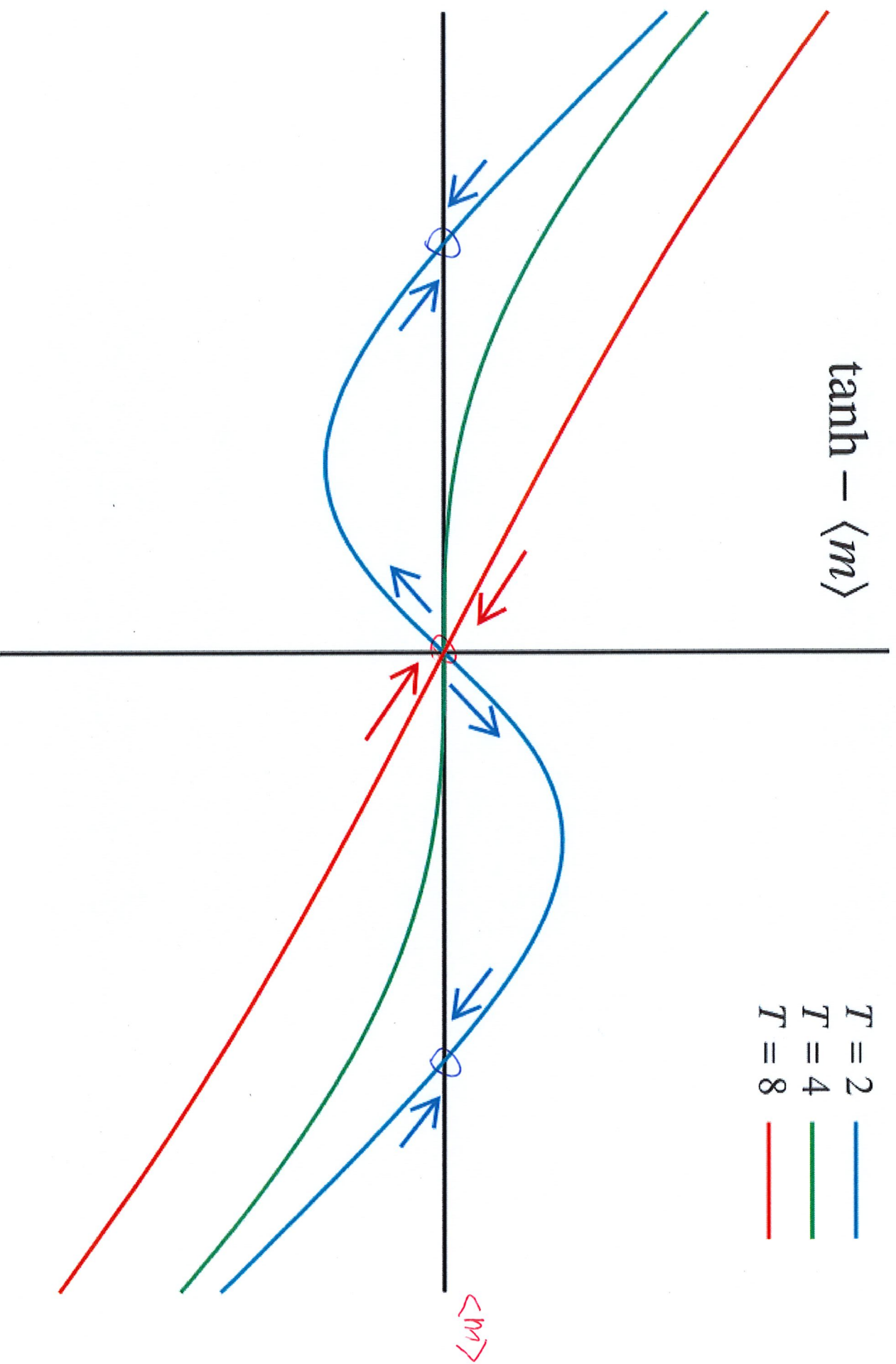
$$\langle m \rangle^2 = 3 \left(\frac{T}{T_c} \right)^3 \left(\frac{T_c}{T} - 1 \right) = 3 \left(\frac{T}{T_c} \right)^2 \left(1 - \frac{T}{T_c} \right)$$

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$\tanh - \langle m \rangle$

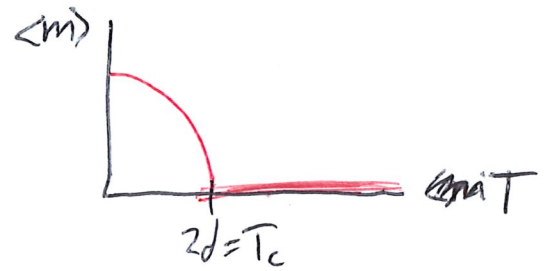
$T = 2$
 $T = 4$
 $T = 8$



For $\frac{T}{T_c} \approx 1 \rightarrow \langle m \rangle \approx \pm \sqrt{3} \left(1 - \frac{T}{T_c} \right)^{1/2} = \pm \sqrt{3} \left(\frac{T_c - T}{T_c} \right)^{1/2}$

$\therefore$ magnetization is continuous at T_c

$$\langle m \rangle \propto \begin{cases} (T_c - T)^{1/2} & \text{for } T \leq T_c \\ 0 & \text{for } T \geq T_c \end{cases}$$



But $\frac{d\langle m \rangle}{dT} \propto \frac{1}{(T_c - T)^{1/2}}$ for $T \leq T_c$

diverges as $T \rightarrow T_c$ from below

$\rightarrow$ mean-field approx predicts 2nd-order PT for Ising model

General result when $m \propto (T_c - T)^b$
with non-integer critical exponent
($b = 1/2$)

Many phase trans. have same sets of critical exponents

$\rightarrow$ universality emergent large-scale behaviour
near critical point
independent of "microscopic" details

Are mean-field predictions correct? ($H=0$ 2nd-order PT $T_c = 2d$
 $b = 1/2$)

$d=1$: No phase transition (Ising 1924) X

$d=2$: 2nd-order phase transition (Onsager 1944) ✓

$$T_c = \frac{2}{\log(1+\sqrt{2})} \approx 2.27 \rightarrow \text{MF } T_c = 4 \text{ off by } \sim 2\times$$

$b = 1/8$, MF off by $4\times$

$d=3$: Numerical analyses $\rightarrow$ 2nd-order $T_c \approx 4.5$ (vs. 6)
 $b \approx 0.32$

$d \geq 4$: $b = 1/2$, $T_c \rightarrow 2d$ as $d \rightarrow \infty$
 $\hookrightarrow$ Mean-field becomes exact

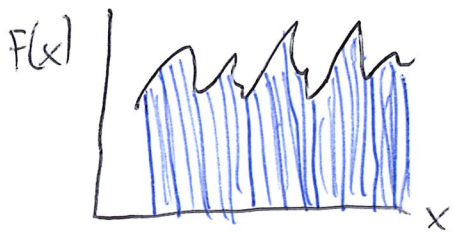
How do we carry out numerical analyses in less than
 $\sim 500,000 \times$ age of universe?
(2d 10×10 part. func.)

Needed for $d \geq 3$ Ising model
and most interacted systems

Sample small subset of micro-states to approximate $\langle O \rangle$

Pseudo-random $\rightarrow$ "Monte Carlo" methods

Example: Compute integral by evaluating integrand at random points



$$\int_{-1}^1 dx \int_{-1}^1 dy H(1 - \{x^2 + y^2\}) = \text{area of disk w/radius } 1 = \pi$$

$$H(r) = \begin{cases} 1 & \text{for } r \geq 0 \\ 0 & \text{for } r < 0 \end{cases}$$

Monte Carlo integration most useful for high-dim'l integrals
like $\langle O \rangle$ for $N \gg 1$ interacting statistical system!

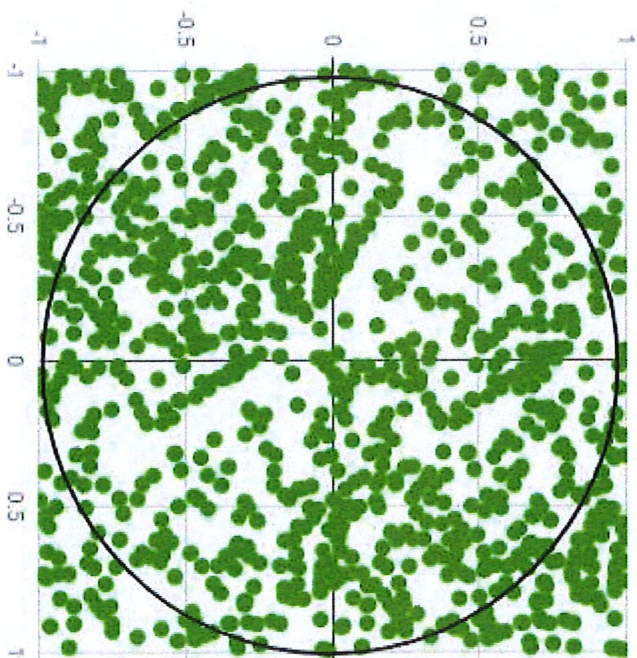
$\frac{\text{length of computation}}{\text{age of universe}} \ll 1 \rightarrow$ very small fraction of micro-states
How can this give reliable approx.

Return to Ising model:

High T : All p_i roughly equal, ~~lets~~

$\langle O \rangle$ determined by degeneracies

Large degeneracy $\rightarrow$ more likely to sample
... could be okay



$$\text{Fraction} = \frac{\pi}{4}$$

Low T : $\langle O \rangle$ dominated ground state
Other contribution suppressed $\sim e^{-E/T}$

Solution: Sample w_i with prob. $P_i \propto e^{-\beta E_i}$
without knowing P_i distribution

Importance sampling algorithms use pseudo-randomness
to find large P_i w/out bias

Example: MTRTT algorithm (1953)

Start from any micro-state

Make a pseudo-random change $\rightarrow \Delta E$

$P_{\text{accept}} = \min \{1, \exp(-\beta \Delta E)\}$ otherwise rejected

$\rightarrow$ New micro-state (possibly unchanged)

Repeat!

Sequence of micro-states is "Markov chain"

(each w_i based on previous one, no memory)

$$\frac{P(A \rightarrow B)}{P(B \rightarrow A)} = \frac{\min \{1, e^{-\beta(E_B - E_A)}\}}{\min \{1, e^{-\beta(E_A - E_B)}\}} = e^{-\beta(E_B - E_A)} = \frac{e^{-\beta E_B}}{e^{-\beta E_A}} = \frac{P_B}{P_A} \checkmark$$

To avoid bias, must (in principle) be able to reach any w_i
from any other w_j

Ergodicity depending on specific system
& pseudo-random update

May take many updates to produce statistically independent w_i
 $\rightarrow$ auto-correlations increase costs, increasing uncertainties
 $\propto 1/\sqrt{N}$ indep. samples

For Ising-~~model~~like systems, huge benefits &

from flipping "cluster" of spins, not just one
(avoid "critical slowing down" near 2nd-order phase trans.
with fluctuation on all length scales)

Example: Lattice quantum field theory

Relativity + QM $\rightarrow$ fields $\Phi(\vec{x}, t)$ fill all space & time

Interactions governed by Lagrangian $\mathcal{L}[\Phi(\vec{x}, t)]$

$\rightarrow$ Feynman path integral

$$Z[\Phi(\vec{x}, t)] = \int D\Phi \exp\left[\frac{i}{\hbar} \int d^3x dt \mathcal{L}[\Phi(\vec{x}, t)]\right]$$

$\sim$ part. func. $T \leftrightarrow \hbar$

$$\hbar t \rightarrow i\tau$$

We have reached open research questions

just a few months after starting

w/probability foundations!