

MATH327: StatMech and Thermo

Wednesday, 7 May 2025

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Something to consider

Suppose we want to *approximately* compute expectation values

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

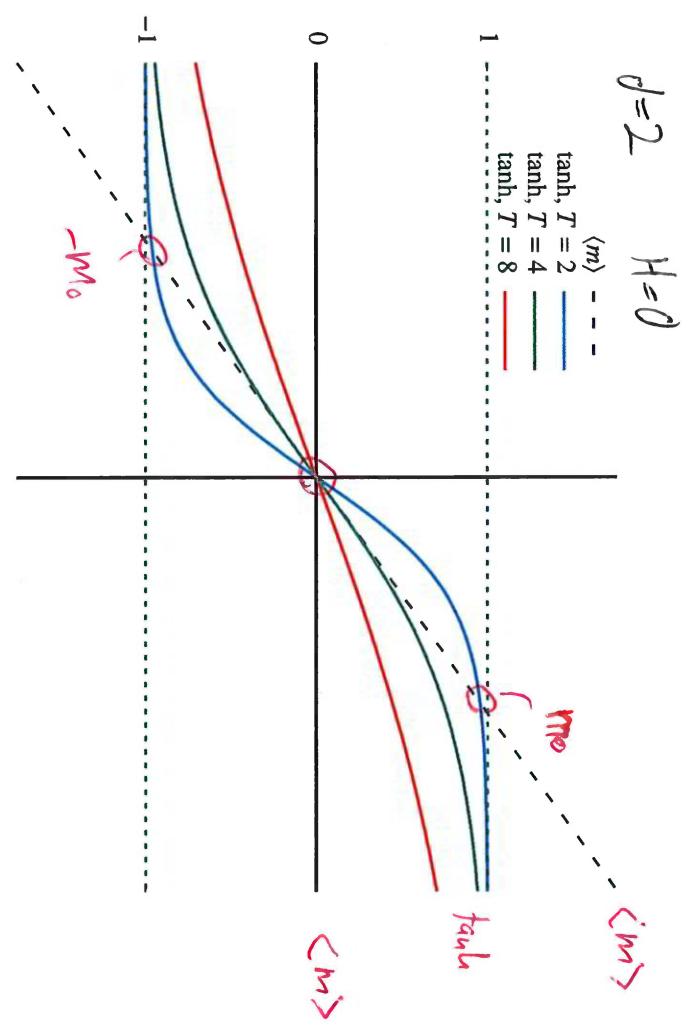
How many of the system's micro-states do we really need to analyse?

Recap
Ising model magnetization as order param.
transition vs. crossover

Mean-field approx. \rightarrow self-consistency condition
 $\langle m \rangle = \tanh(\beta(2d\langle m \rangle + H))$

Plan
Wrap up mean-field approx. for Ising model
Numerical methods - more broadly applicable

Recall mean-field solutions for $d=2$ $T=4$
 $H=0 \rightarrow \langle m \rangle = 0$ (disordered)
 $H \neq 0 \rightarrow \langle m \rangle = \pm m_0 \neq 0$ (ordered)
 $\hookrightarrow m_0 \rightarrow 1$ as $T \rightarrow 0$

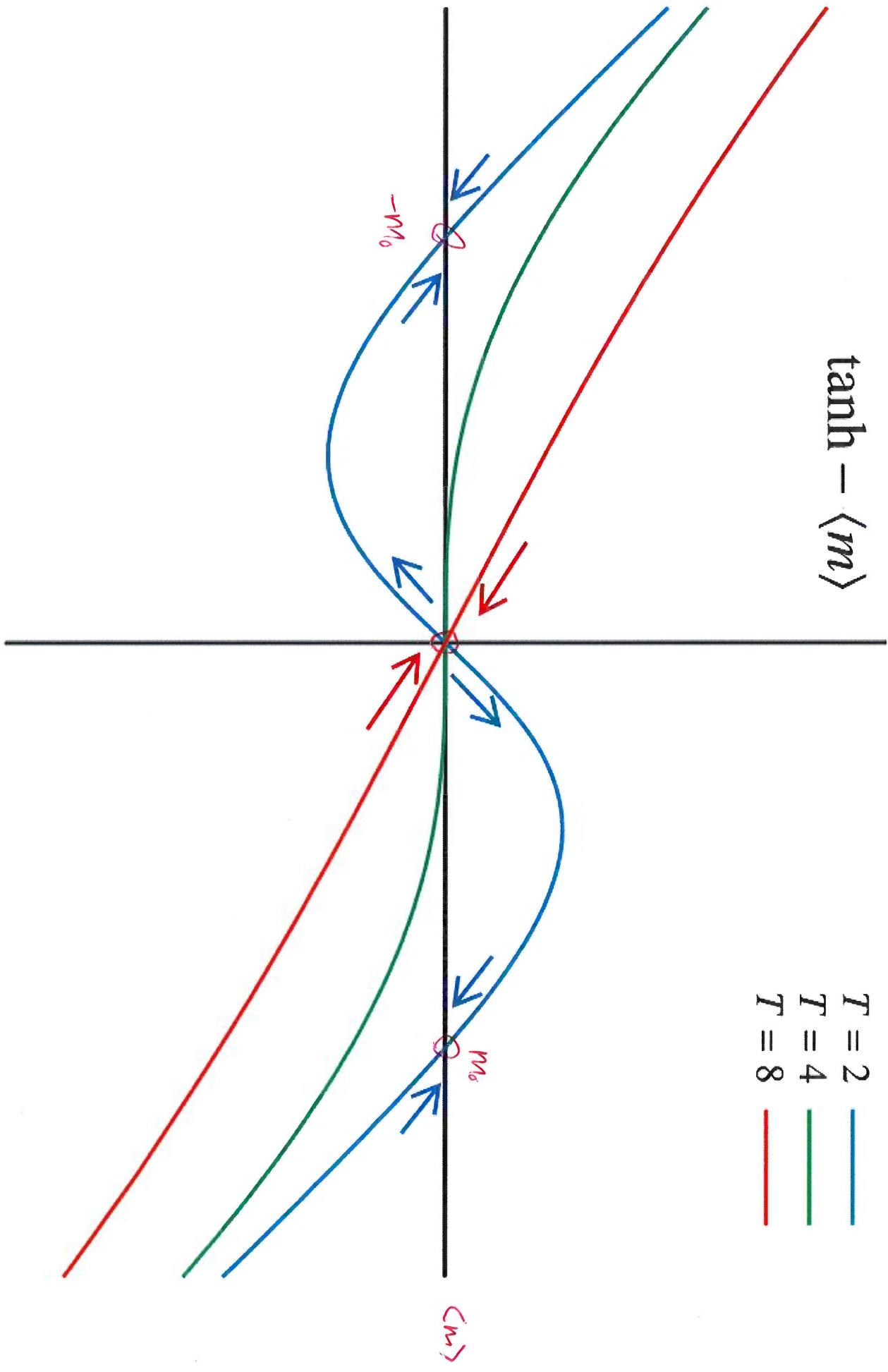


$\tanh - \langle m \rangle$

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



$\langle m \rangle$



Expect low- T ordered phase even with $H=0$

Compare $T=2, 4, 8$

$\langle m \rangle = 0$ always possible

Lower $T \rightarrow$ additional solutions $\langle m \rangle = \pm m_0 \neq 0$
 $m_0 \rightarrow 1$ as $T \rightarrow 0$

which intersection is true solution
adopted by the system?

Consider perturbing $\langle m \rangle = 0 \rightarrow \langle m \rangle = \varepsilon > 0$

$T=8$: $\langle m \rangle$ too large compared to tanh
 \rightarrow decrease to stable $\langle m \rangle = 0$

$T=2$: $\langle m \rangle$ too small

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

Similarly $-\varepsilon \rightarrow -m_0$

Conclude $\langle m \rangle = 0$ unstable \rightarrow ordered phase for low T

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 $H=0$ mean-field approx. give expected phases

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

Need $\tanh > \langle m \rangle$ for $\langle m \rangle = \varepsilon > 0 \rightarrow$ slope > 1 at $\langle m \rangle = 0$

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

Is this a phase trans. w/ discontinuity?

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

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

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

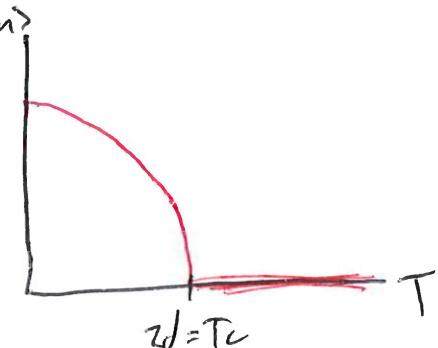
page 141

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

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

So OP continuous:

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



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

diverges as $T \rightarrow T_c$ from below

Mean-field approx. predicts second-order PT for Ising model

occurs whenever $\alpha \propto (T_c - T)^{-b}$

with critical exponent $0 < b < 1$

Many phase trans. have same sets of critical exponents.
→ universality: emergent behaviour near critical point
independent of microscopic details

$H=0$ mean-field predicts second-order PT at $T_c = 2d$, $b = 1/2$

Is this correct?

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

$d=2$: Second-order PT (Onsager 1944) ✓

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

$$b = \frac{1}{8} \rightarrow \text{MF off by } 4x$$

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

$$d \geq 4: b = \frac{1}{2} \checkmark$$

$$T_c \rightarrow 2d \text{ as } d \rightarrow \infty$$

Mean-field becomes exact

(more n.n. \rightarrow more reliable)

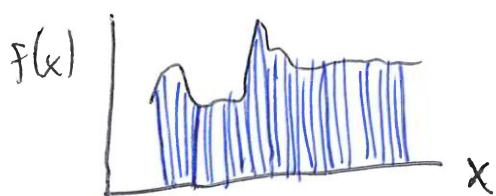
How to do numerical calcs in less than $\sim 500,000 \times$ age of universe?

Needed for most interacting statistical system
 including: $d \geq 3$ Ising model

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

Pseudo-random \rightarrow "Monte Carlo" methods

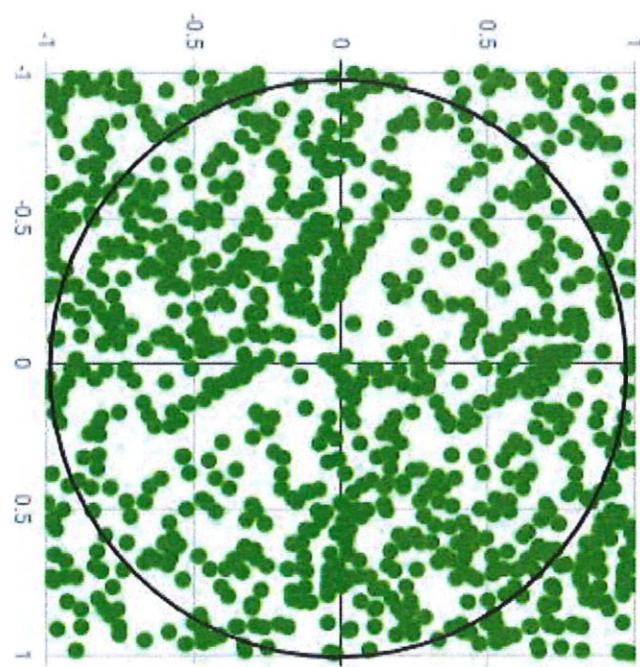
Example: Estimate integral by evaluating integrand at random points



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

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

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



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

length of calc << 1 → very small fraction of micro-states
 age of universe How can this give reliable approx?

High-T Ising model: All p_i roughly equal
 $\langle \sigma \rangle$ determined by degeneracies
 Large degeneracy → more likely to sample...
 could be okay

Low T: $\langle \sigma \rangle$ dominated by ground states
 other contributions 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 without bias

Example: Metropolis-Hastings algorithm (1953)

Start from any micro-state

Pseudo-randomly change $\rightarrow \Delta E$

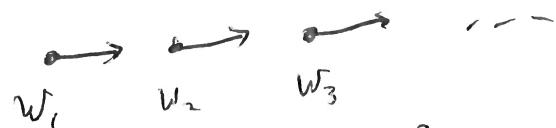
$P_{\text{accept}} = \min \{1, \exp(-\beta \Delta E)\}$, otherwise reject

\rightarrow New micro-state (possibly unchanged)

Repeat

Sequence of micro-states is Markov chain

\rightarrow each w_i based on previous one, no memory



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

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

Ergodicity of random update depending on system

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

For spin systems, huge benefits

from flipping "cluster" of spins, not just one

Reduce "critical slowing down" near 2nd-order phase trans.
with fluctuations on all length scales

We have reached areas of ongoing research
just a few months after starting
w/probability foundations!