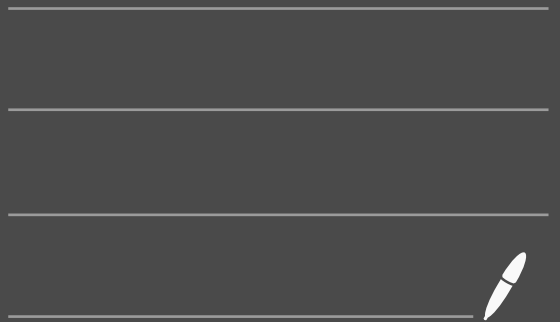
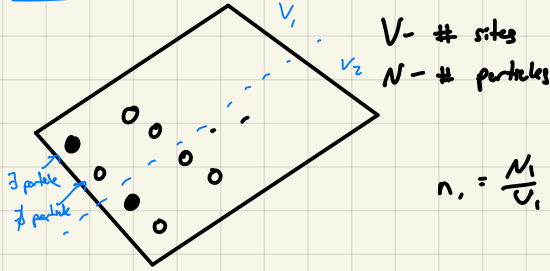


Aizenman



Lecture 1/31 - First day baby!

Adapters



Local density variation

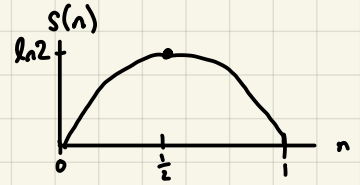
Q: If every configuration is equally likely, what is $\mathbb{P}\{|n_1 - n_2| > \epsilon\}$?

We start by noting that the # of states is $W(N, V) = \binom{V}{N} = \frac{V!}{N!(V-N)!} \approx e^{s(n)V}$

Using the Stirling approx. and a bunch of algebra, $\log N! = N(\ln N - 1) + \ln(2\pi N) + O(\frac{1}{N})$

$$s(n) = -[n \ln n + (1-n) \ln(1-n)]$$

local entropy density



(This is just like Shannon entropy $S(\{p_i\}_{i=1}^N) = -\sum_{i=1}^N p_i \ln p_i$!)

We can say that, for same density difference Δn ,

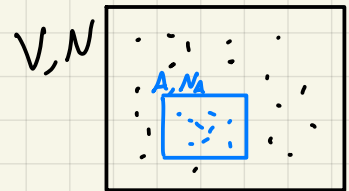
$$\mathbb{P}\{n_1 - n_2 = \Delta n\} = \frac{W(V_1, N_1) W(V_2, N_2)}{W(V, N)} = \frac{e^{[s(n + \frac{\Delta n}{2}) + s(n - \frac{\Delta n}{2})]V/2}}{e^{s(n)V}}$$

Taylor 1st order $\approx e^{-\frac{1}{2} s''(n) (\frac{\Delta n}{2})^2 \frac{V}{2}}$ very small for large volume!

equivalence of ensemble

Q: If we have a subvolume A, what is the distribution of the # of particles in A?

It turns out that it follows a Poisson distribution that is identical for different A's and depends only on $|A|$.



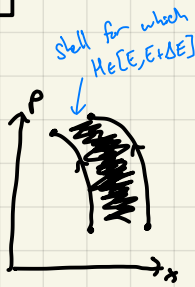
Stat Mech Setup

On the macroscopic level:

- divisibility
- additivity of V, N, E, \dots
- many DOFs (spin states of constituents, etc.)
- a natural notion of counting microstates

examples of microstate counting

Recall from PHY205 that phase space evolutions of $((\vec{x}_1, \dots, \vec{x}_N), (\vec{p}_1, \dots, \vec{p}_N))$ preserve the Liouville measure $\int dx \dots dp \dots$, \Leftrightarrow the volume in phase space \Leftrightarrow # of states $\Rightarrow W(V, N) = \int_V \int_{\mathbb{R}^{2N}} \mathbb{1}\{H_N(\vec{x}, \vec{p}) \in (E, E + \Delta E)\} d\vec{x} d\vec{p}$



PHY208 would say $W(V, N) = \text{Tr } \rho_{(E, E+\Delta E)}$ free of density matrix

Note: Since s is convex and $S = e^{sV}$, S is also convex.

So, variational formulations of stat mech. allow states that maximize the convex objective S .

We arrive at the fact that

$$W([E, E+\Delta E]) \propto e^{S\left(\frac{E}{V}\right) \cdot |V|}$$

local
entropy density
of energy
density

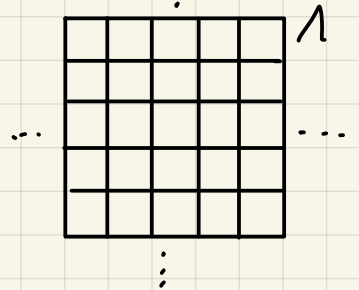
Lecture 2/2 - Partition Fns + Ensembles

We consider both discrete & continuous models. We describe the configuration of a model by defining on a domain G , which is often a lattice. At each site, we have possible values that depend on the model. More formally, $w: G \rightarrow \Omega$, where

$\Omega = \{0, 1\}$
adsorption

$\Omega = \{-1, 1\}$
Ising model

$\Omega = \mathbb{R}^d$
continuum



Partition Functions

For a given space, we define the partition function by

$$Z_N(\beta) = \int_{\Lambda^N} \int_{\mathbb{R}^{d \cdot N}} e^{-\beta H_N(\vec{x}, \vec{p})} \prod_{j=1}^N dx_j d p_j$$

Liouville Measure preserved during the evolution

The "canonical ensemble" allows us to not exclude configurations, but to fine-tune the desirable configs. space via β (or chemical potential μ if we use $e^{-\beta H}$)

$$= \int_{\mathbb{R}} e^{-\beta E} e^{S(E)} dE \approx |\Lambda| \int_{\mathbb{R}} e^{-\beta u |\Lambda|} e^{s(u) |\Lambda|} du$$

$$\approx |\Lambda| \int_{\mathbb{R}} e^{[s(u) - \beta u] |\Lambda|} du$$

(When we're in a discrete model, we define Z_N as a discrete sum over the discrete phase space)

$$Z_N(\beta) = \sum_{w \in \Omega} e^{-\beta H_N(w)}$$

↑ energy of certain cell
↑ inverse temp

In such integrals, since u is normally smooth + bounded, we expect it to be dominated by $\sup_u (s(u) - \beta u)$.

Assuming entropy actually behaves as $W([E, E+\Delta E]) \approx e^{s(\frac{E}{|\Lambda|)} \cdot |\Lambda|}$, then

$$\lim_{|\Lambda| \rightarrow \infty} \frac{1}{|\Lambda|} \ln Z_N(\beta) = \sup_u \{s(u) - \beta u\}$$

Legendre transform of $s(\cdot)$

"winner takes all"

Types of ensembles

"microcanonical"

$\cdot H_N \in [E, E+\Delta E]$

"canonical"

$\cdot e^{-\beta H_N}$
 \cdot releases bounds on E

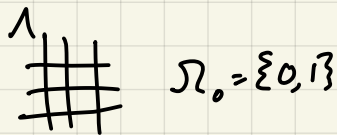
"grand canonical"

$\cdot e^{\beta[-H + \mu N + hM]}$

↑ energy ↑ # particles ↑ magnetization

\cdot releases bounds on other extensive properties like N, M

Example: Adsorption on a lattice



$$\Omega_0 = \{0, 1\}$$

$$Z_n(\mu) = \sum_{w \in \Omega_n} e^{\mu N(w)} = \sum_{w \in \Omega_n} \prod_i e^{\mu w_i} = (1 + e^\mu)^{|n|}$$

of particles in config.
prod. over sites

sum over all configurations

Let $n = \frac{N}{|n|}$ be the particle density. Then,

$$\sup_n \{s(n) - \mu(n)\} = \lim_{|n| \rightarrow \infty} \frac{1}{|n|} \ln (1 + e^\mu)^{|n|} = \ln(1 + e^\mu)$$

Legendre transform of $s(\cdot)$

This matches the result $s(n) = -n \ln(n) - (1-n) \ln(1-n)$ that we found for adsorption via the Stirling approx.

Convexity

Def: A set $D \subset \mathbb{R}^2$ is **convex** if $\forall x, y \in D$,

$$tx + (1-t)y \in D \quad \forall t \in [0, 1].$$

D contains line between x and y

A function $f: D \rightarrow \mathbb{R}$ is **convex** if $\forall x, y \in D$,

$$f(tx + (1-t)y) \leq tf(x) + (1-t)f(y)$$

f lies below line between $f(x)$ and $f(y)$

Equivalently, $\forall x \in D$ and a fixed $y \in D$, $\frac{f(y) - f(x)}{y - x}$ is monotone increasing.

If f is twice differentiable, a sufficient condition of convexity is

$$f''(x) \geq 0 \quad \forall x \in D.$$

Theorem: Let $\{G_\alpha(x)\}_\alpha$ be a family of linear functions of x . Then,

$$F(x) = \sup_\alpha G_\alpha(x) \quad \text{is convex.}$$

Proof: Intersection of closed half-spaces, which are all convex, is itself convex.

D

Note that the Legendre Transform looks similar: it is indeed the case that Legendre transforms are convex.

Theorem: \forall convex $F: [a, b] \rightarrow \mathbb{R}$,

① F is differentiable everywhere except at a countable number of points.

② On the set where $F'(x)$ exists, it is monotone increasing.

③ $\forall x \in [a, b]$, both $F'_-(x) = \lim_{\varepsilon \rightarrow 0} \frac{F(x-\varepsilon) - F(x)}{-\varepsilon}$ and $F'_+(x) = \lim_{\varepsilon \rightarrow 0} \frac{F(x+\varepsilon) - F(x)}{\varepsilon}$
exist and satisfy $\lim_{\varepsilon \rightarrow 0} F'_\pm(x-\varepsilon) \leq F'_-(x) \leq F'_+(x) \leq \lim_{\varepsilon \rightarrow 0} F'_\pm(x+\varepsilon)$

④ The right derivatives are continuous a.e., and where this happens, F' exists.

Lecture 2/7 - Convexity + Legendre Transform

Def. The **Legendre Transform** of a convex function F is

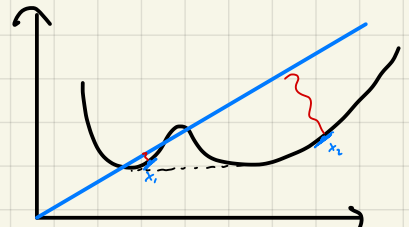
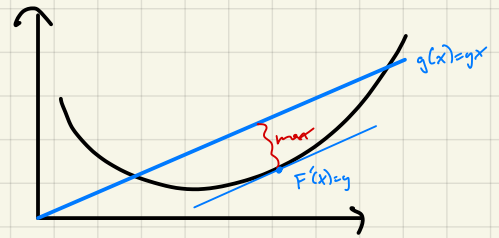
$$(TF)(y) = \sup_x \{ y \cdot x - F(x) \}$$

← inf for concave

In a sense, varying y explores the values of F for which F' takes the value y .

The transform T is itself convex, since it is the max of linear functions.

If F is not convex, T computes the Legendre Transform of the convex hull of F .



Both points have same F' , but x_2 is selected by the sup.

Theorem: (Inverted Property of the Legendre Transform)

$$\forall \text{ convex } F: \mathbb{R} \rightarrow \mathbb{R}, \quad T(TF) = F$$

Proof: We prove this assuming F differentiable, but the result holds generally.

Let $x(y)$ be the point where $F'(x)=y$. Then,

$$(TF)(y) = y \cdot x(y) - G(x(y)) \quad \text{and} \quad G'(x(y)) = y$$

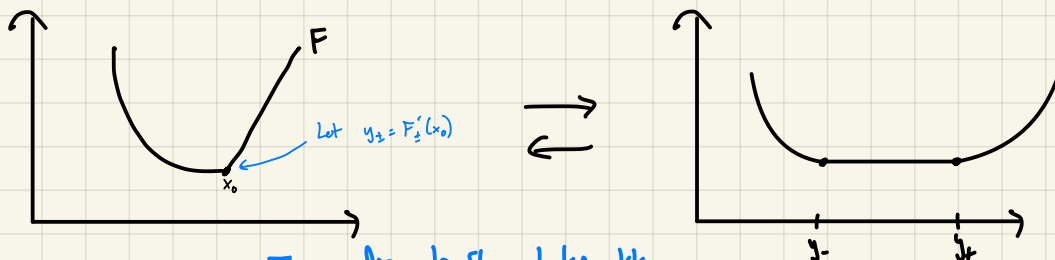
$$\text{So, } (T^2F)(x) = \sup_y \{ xy - (TF)(y) \} = x \cdot y(x) - (TF)(y(x)) \quad \text{for } y(x) \text{ s.t. } (TF)'(y) = x$$

$$\text{We can compute } (TF)'(y) = x(y) + y \cdot x'(y) - \frac{G'(x(y)) \cdot x'(y)}{1} = x(y)$$

$$\Rightarrow (T^2F)(x) = x \cdot y(x) - y(x) \cdot x(y(x)) + G(x(y(x))) = G(x).$$

□

If we have a F with a kink, Legendre Transform maps the kink to a flat region, and vice versa.



This discontinuity looks like a first-order phase transition.

Def: An invertible **measure-preserving transformation** $T: \Omega \rightarrow \Omega$ satisfies $w \mapsto T_w$

$$\mu(T^{-1}(E)) = \mu(E) \quad \forall E \subseteq \Omega \text{ measurable.}$$

they differ by a set of measure 0!

Def: A measure-preserving transformation T is **ergodic** if $f(Tx) \stackrel{a.s.}{=} f(x)$ holds only for constant f . In other words, if E is s.t. $T^{-1}(E)$ and E differ by a set of measure 0, then either $\mu(E) = 0$ or $\mu(E^c) = 0$.

Thm: (Birkhoff)

If $\{T_i\}_{i \in \mathbb{N}}$ is a collection of ergodic, measure-preserving transformations, then \forall bounded, measurable $f: \Omega \rightarrow \mathbb{R}$, the limit exists and equals

$$\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N f(T^i x) \stackrel{a.e.}{=} \int_{\Omega} f(x) \mu(dx)$$

If $\{T_t\}_{t \geq 0}$ is a collection of ergodic, measure-preserving transformations, then \forall bounded, measurable $f: \Omega \rightarrow \mathbb{R}$, the limit exists and equals

$$\lim_{\tau \rightarrow \infty} \frac{1}{\tau} \int_0^{\tau} f(T_t) dt \stackrel{a.e.}{=} \int_{\Omega} f(x) \mu(dx)$$

Time averages are equivalent to probability averages!

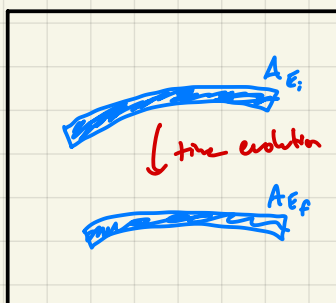
The **ergodic hypothesis** is, in general, not true. Not all microstates are equally probable after a long time, as we see below.

Thm: (Poincaré Recurrence)

Let T be a measure-preserving transformation on $(\Omega, \mathcal{B}, \mu)$ and $A \subseteq \Omega$ measurable with $\mu(A) > 0$. Then, for a.e. $x \in A$, $\exists m \in \mathbb{N}$ s.t. $T^m x \in A$.

For sets of pos. measure, we eventually return to the set.

Boltzmann statistics is based on the **equidistribution hypothesis**:



$$A_E = \{w \mid E \leq H(w) \leq E + \Delta E\} \quad \text{where} \quad \Delta E = o(E) \text{ is tiny}$$

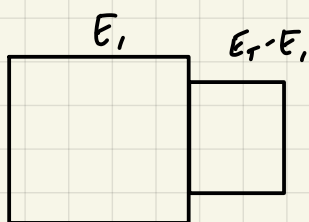
We know that under Hamiltonian mechanics, Liouville measure is preserved.

only if that's robustly preserved by the evolution

The **equilibrium hypothesis** states that the system equilibrates to a distribution in Ω with same Liouville measure and the correct E .

Thermodynamics

Fun fact: gravity is not thermodynamically stable



We have two systems in equilibrium states with total energy E_r . When we combine the two systems, the equilibrium energy is such that the total entropy

$$S = \underbrace{k_B \log W_1(E_1)}_{S_1(E_1)} + \underbrace{k_B \log W_2(E_r - E_1)}_{S_2(E_r - E_1)} \quad \leftarrow \text{entropy is additive}$$

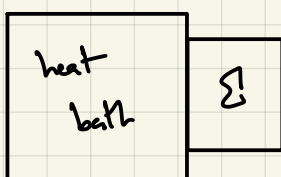
is maximized over E_1 .

$$\Rightarrow \text{in equilibrium, } \left. \frac{dS_1}{dE_1} \right|_{E_1} = \left. \frac{dS_2}{dE_1} \right|_{E_r - E_1}$$

Since we know equality of temperature is the condition for equilibrium, we define the temperature T , in Kelvin, to be

$$\frac{\partial S}{\partial E} = \beta = \frac{1}{k_B T}$$

Consider a heat bath with constant temperature T , and a system Σ .



The variational principle maximizes w.r.t. E_Σ

$$S_\Sigma(V, E_\Sigma, N, \dots) + S_{\text{Bath}}(E_{\text{Tot}} - E_\Sigma)$$

Since T & β constant for the bath, this equals

$$\approx S_\Sigma(V, E_\Sigma, N, \dots) + \underbrace{S_{\text{Bath}}(E_{\text{Tot}})}_{\text{constant}} - \beta E_\Sigma$$

So, we in effect maximize Legendre transform!

$$-\beta F(V, \beta, N) = \sup_{E_\Sigma} \left[S_\Sigma(V, E_\Sigma, N) - \beta E_\Sigma \right]$$

F is the Helmholtz free energy, and E is the available energy when held to a constant temp. β . So, β is the Legendre transform dual to energy!

If we hold T and P fixed, the thermodynamic potential is

$$-\beta G(p, T, N_1, \dots, N_r) = -\frac{\beta}{k_B} \inf_E \left[E - TS(V, E, N) \right]$$

$$\Rightarrow G(p, T, \mu) = \inf_{E, V} \left[E + pV + \sum_j \mu_j N_j - TS(V, E, N_1, N_2, \dots) \right]$$

This is the Gibbs free energy.

Inverting the Legendre Transform yields

$$S(V, E, N) = \frac{-\partial G(T, p, N)}{\partial T}$$

Places where G has a kink singularity correspond to **first-order phase transitions**.

Lecture 2/14 -

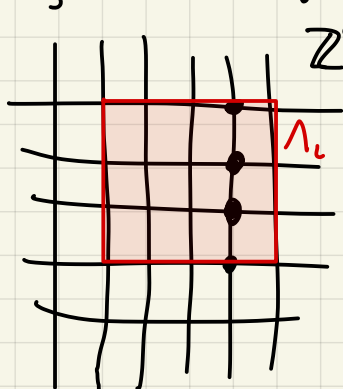
From the definition of $G(\cdot)$, we can write the differential form as

$$dG = -SdT + Vdp + \sum_j \mu_j dN_j$$

Statistical Mechanics

We would like to investigate the nature of entropy.

We work with finite-dim graph (meaning as size diverges, the size of boundary is $o(\text{volume})$), say \mathbb{Z}^d . This graph is homogeneous to translation, and is tileable (by cubes in this case).



\mathbb{Z}^d $\{\sigma_x\}_{x \in \mathbb{Z}^d}$ - each lattice point

Ω_0 - measurable set of outcomes for each σ_x

$\mu_0(d\sigma)$ - measure on outcomes in Ω_0

$\Omega_A = \Omega_0^A = \{\omega: A \rightarrow \Omega_0\}$ - all possible lattice configurations of $A \subset \mathbb{Z}^d$

$\omega = (\sigma_x)_{x \in \mathbb{Z}^d}$

$\Lambda_L = \{1, \dots, L\}^d$ - linear box of size L

(In the example of the Ising model, a single spin can take values $-1, 0, 1$ with equal probability. In this case, $\Omega_0 = \{-1, 0, 1\}$, and μ assigns equal weight.)

We turn to the **extensive energy function**, also called the Hamiltonian.

As an example, in the Ising model, $\sigma_u \in \{-1, 1\}$ and

$$H_\Lambda(\omega) = - \sum_{(u,v) \in \Lambda^2} J_{u,v} \sigma_u \sigma_v - h \sum_{u \in \Lambda} \sigma_u$$

↑ interacting term coupling constant
↑ external field orientation

This is an example where energy is given to pairs and singlets.

More generally,

$$H_\Lambda(\vec{\sigma}) = \sum_{A \subset \Lambda} J_A \Phi_A(\vec{\sigma}_A)$$

is a framework to describe interactions among all subsets.

We can easily bound by

$$\max_{w \in \Omega_\Lambda} |H_\Lambda(w)| \leq \sum_{x \in \Lambda} \sum_{A \ni x} \frac{1}{|A|} |J_A| \max_{\sigma_A} |\Phi_A(\sigma_A)|$$

Infinity norm of Φ_A .
We assume this = 1 st. J is a good choice of coupling

$$\Rightarrow \|J\| = \sum_{A \ni 0} \frac{1}{|A|} |J_A|$$

and

$$\max_{w \in \Omega_\Lambda} |H_\Lambda(w)| \leq |\Lambda| \|J\|$$

Now, let us work out the partition function

$$Z_\Lambda(\beta, J) = \int_{\Omega_\Lambda} e^{-\beta H_\Lambda(\sigma)} \mu(d\sigma)$$

Taking the thermodynamic limit $|\Lambda| \gg 1$, we split it into a sum of slices of configuration space sharing the same energy $E = u |\Lambda|$, where u is the energy density.

$$\Rightarrow Z_\Lambda(\beta) \approx \int e^{-\beta u |\Lambda|} e^{S(u) |\Lambda|} du = \int e^{[-\beta u + S(u)] |\Lambda|} du$$

$$\Rightarrow \frac{1}{|\Lambda|} \log Z_\Lambda \xrightarrow{|\Lambda| \rightarrow \infty} \max_u [S(u) - \beta u] = -\beta F(\beta)$$

can be thought of as thermodynamic pressure

Legendre transform
because rate of exponential growth dominates the integral

free energy

Theorem: (Existence of Pressure Function)

$$J_A = \sum_{A \ni u} J_{A \cup u} \text{ and } \Phi_{A \cup u}(\sigma_x) = \Phi_A(\sigma_{x-u})$$

For any translation-invariant system, the following limit exists

$$\Psi_\Lambda(\beta, J) = \lim_{L \rightarrow \infty} \frac{1}{|\Lambda_L|} \log Z_{\Lambda_L}(\beta) \quad (\|J\| < \infty)$$

Important: this links the stat mech construction to things that are useful for thermo!

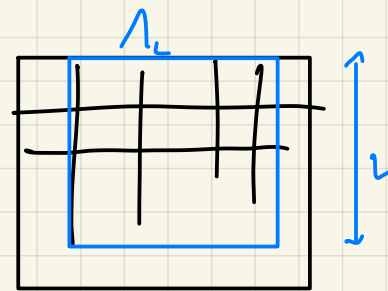
Lecture 2/16-

Pressure Function in Thermodynamic Limit

The object of interest is

$$Z_{\Lambda_L}(\beta, \dots) = \int_{\Omega_{\Lambda_L}} e^{-\beta H_{\Lambda_L}(\omega)} \mu(d\omega)$$

microstates



Letting

$$\Psi(\beta, \dots) = \lim_{L \rightarrow \infty} \frac{1}{|\Lambda_L|} \log Z_{\Lambda_L} \quad (*)$$

we see a "winner takes all" principle, where the ω with the smallest H is most likely to be observed.

We prove that the thermodynamic limit of (*) exists with a box-chopping argument: draw smaller boxes of size k , ignore interaction terms along the boundaries such that energy is additive and μ is multiplicative. Taking $k, L \rightarrow \infty$ together, the surface area ratio of the k -boxes converges to 0, and so the sequence

$$\left\{ \frac{1}{|\Lambda_L|} \log Z_{\Lambda_L} \right\}_L$$

is Cauchy and has a limit.

(In fact, $\frac{1}{|\Lambda_L|} \log Z_{\Lambda_L} \approx \frac{1}{|\Lambda_k|} \log Z_{\Lambda_k} + O(\frac{1}{k})$)

surface vol.

This only works for Hamiltonians with short range interactions across the boundary. For the general case, it helps to truncate long-range interactions and bound the error.

Let $\|J\| = \sum_{A \ni 0} \frac{1}{|A|} J_A$ and $J_A^{(R)} = J_A \cdot \mathbb{1}_{\{\text{diam}(A) \leq R\}}$

truncated to R

We wish to also prove the existence of the thermodynamic limit for $J^{(R)}$.

Now,

$$\left| H_{\Lambda}(\omega) - H_{\Lambda}^{(R)}(\omega) \right| \leq \left| \sum_x \sum_{\substack{A \subset \Lambda \\ \text{diam}(A) > R \\ A \ni x}} \frac{1}{|A|} J_A \phi_A(\omega_A) \right| \leq |\Lambda| \underbrace{\|J - J^{(R)}\|}_{\leq \epsilon}$$

If $\|J\| < \infty$, then $\forall \epsilon > 0 \exists R > 0$ s.t. $\sum_{\substack{A \ni x \\ \text{diam}(A) > R}} \frac{1}{|A|} |J_A| < \epsilon$ (bound the tail w.r.t. R)

So,

$$Z_n = \int_{\Omega_n} e^{-\beta H_n^{(R)}(\omega) - \beta \frac{(H_n - H_n^{(R)})}{\varepsilon |\Lambda|}(\omega)} \mu(d\omega)$$

$$\Rightarrow e^{-\beta \varepsilon |\Lambda|} Z_n^{(R)} \leq Z_n \leq e^{\beta \varepsilon |\Lambda|} Z_n^{(R)}$$

$$\Rightarrow \left| \frac{1}{|\Lambda|} \log Z_n^{(R)} - \frac{1}{|\Lambda|} \log Z_n \right| \leq \beta \varepsilon$$

Since $\left\{ \frac{1}{|\Lambda|} \log Z_n \right\}_L$ converges and we can arbitrarily approximate with large enough R , then the finite-interaction approx. also converges!

In fact, we can bound the distance between these limits by

$$\left| \Psi(\beta, J) - \Psi(\beta, J^{(R)}) \right| \leq \beta \|J - J^{(R)}\|$$

$$\Rightarrow \Psi(\beta, J) = \lim_{R \rightarrow \infty} \Psi(\beta, J^{(R)})$$

However, since the Ψ 's are convex, the PSET problem 3.1 reveals that the derivatives also converge. This is a very useful property because we can view Ψ as a generating function!

mean of H over states

$$\frac{d}{d\beta} \Psi_n(\beta, \dots) = \frac{-1}{|\Lambda|} \frac{\int H_n(\omega) e^{-\beta H_n(\omega)} \mu(d\omega)}{\int e^{-\beta H_n(\omega)} \mu(d\omega)} = \frac{1}{|\Lambda|} \langle H \rangle_{n,\beta}$$

variance of H over states

$$\frac{d^2}{d\beta^2} \Psi_n(\beta, \dots) = \dots = \frac{1}{|\Lambda|} \left(\langle H^2 \rangle_{n,\beta} - \langle H \rangle_{n,\beta}^2 \right) = \frac{1}{|\Lambda|} \langle [H - \langle H \rangle]^2 \rangle$$

In general, $\Psi_n(\beta, \dots) = \frac{1}{|\Lambda|} \log \int e^{-\beta H_n(\omega)} \mu(d\omega)$
is the cumulant generating function of H_n !

This yields several properties of Ψ !

① $\Psi_n(\beta)$ is convex in β ($\Psi_n'' \geq 0$ since variance ≥ 0)

② $\Psi(\beta)$ is convex in β (pointwise limit is convex)

③ At a.e. β , $\Psi(\beta)$ is differentiable and $\lim_{L \rightarrow \infty} \frac{\langle H_n \rangle_{n,\beta}}{|\Lambda|} = \frac{d}{d\beta} \Psi(\beta)$

Also, $\int_{\beta_1}^{\beta_2} \Psi''(\beta) d\beta = \Psi'(\beta_2) - \Psi'(\beta_1) \Rightarrow m \left(\left\{ \beta: \left\langle \left(\frac{H_n - \langle H_n \rangle}{\sqrt{n}} \right)^2 \right\rangle > b \right\} \right) \leq \frac{\beta_2 - \beta_1}{b}$

So, the regions where Ψ'' is large are rather small.

Gibbs Equilibrium States

Def: We have **microstates** ω , which are classically configurations in our configuration space Ω and quantumly vectors in the Hilbert space.

Def: **Observables** are classically functions $F(\omega)$ over Ω and quantumly are operators on our Hilbert space.

Def: **States** are expectation-value functionals

$$\rho: F \rightarrow \langle F \rangle_\rho, \quad F \mapsto \int_{\Omega} F(\omega) \rho(d\omega)$$

given by a probability measure on Ω .

So, $\langle F \rangle_\beta = \int_{\Omega} F(\omega) \underbrace{e^{-\beta H_n(\omega)} \mu(d\omega)}_{\text{Gibbs measure } \Delta(d\omega)} \underbrace{\frac{1}{Z_n(\beta)}}_{\text{is the expectation in the Gibbs canonical ensemble.}}$

The measure $\Delta(d\omega) = \frac{e^{-\beta H(\omega)} \mu(d\omega)}{Z(\beta)}$ is a tilted version of the a priori distribution μ over configuration space.

We can generalize this "tilting" via the following measure theory lingo.

Def: Given a (finite) measure space $(\Omega, \mathcal{B}, \mu)$ and a $f: \Omega \rightarrow \mathbb{R}$ that is normalized ($\int f(\omega) \mu(d\omega) = 1$), then

$\Delta(d\omega) = f(\omega) \mu(d\omega)$ is a measure and $f = \frac{\delta \Delta}{\delta \mu}$ is the **Raden-Nikodym derivative**.

Furthermore, the **entropy** of ρ over μ is given by

$$S(\rho | \mu) = - \int_{\Omega} f(\omega) \log(f(\omega)) \mu(d\omega) = - \int \log(f(\omega)) \Delta(d\omega)$$

Recall the general Jensen's inequality:

Theorem (Probability Jensen):

For a measure space (X, \mathcal{B}, μ) of positive measure, any integrable $g: X \rightarrow \mathbb{R}$, and any concave $F: \mathbb{R} \rightarrow \mathbb{R}$,

$$\int F(g(x)) \mu(dx) \leq F(\langle g \rangle_\mu) \mu(X)$$

where $\langle g \rangle_\mu$ is the normalized mean of g and is given by

$$\int_X \langle g \rangle_\mu \mu(dx) = \int_X g(x) \mu(dx)$$

With this, we can prove:

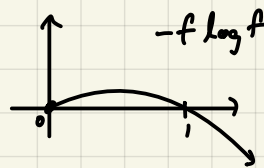
Theorem:

$$S(\rho|\mu) \leq 0, \quad \text{with equality iff } f(\omega) = \frac{1}{\int \mu(d\omega)}$$

Proof: The Jensen inequality on $g(f) = -f \log f$ leads

$$S(\rho|\mu) = \int g(f(\omega)) \mu(d\omega) \leq g\left(\int f(\omega) \mu(d\omega)\right) = 0$$

$\leftarrow g(1) = 0$



The bound $S(\rho|\mu) \leq 0$ (in fact $\leq \log \mu(\Omega_n)$ for unnormalized) yields a **variational characterization of Gibbs states.**

Theorem: (Variational Gibbs)

For a finite system with a-priori measure μ and energy function $H_n(\omega)$, the Gibbs measure $\rho_{n,\beta}(d\omega) = \frac{e^{-\beta H_n(\omega)} \mu(d\omega)}{Z_n(\beta)}$ minimizes the state function

$$F(\rho) := \int_{\Omega_n} H_n(\omega) \rho(d\omega) - \frac{1}{\beta} S(\rho|\mu) = \langle H_n \rangle_\rho - \frac{1}{\beta} S(\rho|\mu)$$

Note that β controls the weight of the energy minimization and entropy maximization.

Large β (small T) prefer ground states while small β (high T) prefer high entropy!

Pressure as Gibbs Measure's Generating Function

Recall the **pressure function** given by

$$\Psi_n(\beta, \mathcal{J}) := \frac{1}{|A|} \log z_n(\beta, \mathcal{J})$$

We saw that

$$\frac{\partial}{\partial \beta} \Psi_n(\beta, \mathcal{J}) = -\frac{1}{|A|} \langle H_n \rangle_{\beta, \mathcal{J}} \quad \text{and} \quad \frac{\partial^2}{\partial \beta^2} \Psi_n(\beta, \mathcal{J}) = \frac{1}{|A|} \text{Var}(H_n)_{\beta, \mathcal{J}} \geq 0$$

More generally, $\forall A \subseteq \Lambda$ we have

$$\frac{\partial}{\partial \mathcal{J}_A} \Psi_n(\beta, \mathcal{J}) = -\frac{\beta}{|A|} \left\langle \frac{\partial}{\partial \mathcal{J}_A} H_n \right\rangle_{\beta, \mathcal{J}}$$

Convexity arguments can also give

$$\lim_{L \rightarrow \infty} \frac{1}{|A_L|} \langle H_{A_L} \rangle_{\beta, \mathcal{J}} = -\frac{\partial}{\partial \beta} \Psi(\beta, \mathcal{J}) \quad \text{"limiting expected energy density" is derivative of } \Psi(\beta)$$

This holds true for all choices of b.c.'s for the finite volumes Λ_L .

However, for β at which $\frac{\partial}{\partial \beta} \Psi(\beta, \mathcal{J})$ is discontinuous, values of $\frac{1}{|A_L|} \langle H_{A_L} \rangle_{\beta, \mathcal{J}}$ depend on boundary conditions, yielding a **first-order phase transition!**

For such β 's, the range of observable energy densities collapses as $L \rightarrow \infty$ to the interval

$$- \left[\frac{\partial}{\partial \beta} \Psi(\beta+0, \mathcal{J}), \frac{\partial}{\partial \beta} \Psi(\beta-0, \mathcal{J}) \right]$$

Concentration of Measure

(google: (1) Cramér large deviation expansion for martingales)
(2) Doob-Vadhan theory of large deviations)

Theorem: (Concentration of energy density)

For any extensive system with Hamiltonian of the form

$$H_\Lambda(\phi) = \sum_{A \subseteq \Lambda} J_A \Phi_A(\phi) = \sum_{x \in \Lambda} \left(\sum_{A \ni x} \frac{1}{|A|} J_A \Phi_A(\phi) \right)$$

for each $\beta < \infty$ there are functions of the form $\delta_{\beta, \pm}$ st. $\forall \epsilon > 0$, at large enough L we have

\mathbb{P} over Gibbs-filled α -priori measure

$$\mathbb{P}_{\beta, \Lambda}^{\#} \left\{ \frac{1}{|\Lambda|} H_{\Lambda}^{\#} \leq \frac{-\partial \Psi}{\partial \beta}(\beta + 0, J) - \epsilon \right\} \leq e^{-\delta_{\beta, +}(\epsilon) |\Lambda|}$$
$$\mathbb{P}_{\beta, \Lambda}^{\#} \left\{ \frac{1}{|\Lambda|} H_{\Lambda}^{\#} \geq \frac{-\partial \Psi}{\partial \beta}(\beta - 0, J) + \epsilon \right\} \leq e^{-\delta_{\beta, -}(\epsilon) |\Lambda|}$$

"probability of energy density deviation is exponentially small in volume"

Lecture 2/23-

Recap

Recall Gibbs states given by measure $\Delta_{\lambda, \beta}(d\omega) = \frac{e^{-\beta H_{\lambda}(\omega)} \mu_{\lambda}(d\omega)}{Z_{\lambda}(\beta, \dots)}$

If we define **Free Energy** to be

$$F_{\beta}(\lambda) = \int_{\Omega} H_{\lambda}(\omega) \Delta(d\omega) - \frac{1}{\beta} S(\Delta \| \mu) = \langle H_{\lambda} \rangle_{\Delta} - \frac{1}{\beta} S(\Delta \| \mu)$$

A minimizer of F minimizes $\langle H_{\lambda} \rangle_{\Delta} - \frac{1}{\beta} S(\Delta \| \mu)$, or equivalently it maximizes

$$\begin{aligned} S(\Delta \| \mu) - \beta \langle H_{\lambda} \rangle_{\Delta} &= - \int_{\Omega} \log \left(\frac{\Delta_{\lambda}(\omega)}{\mu_{\lambda}(\omega)} \right) \Delta_{\lambda}(\omega) \mu(d\omega) - \langle H_{\lambda} \rangle_{\Delta} \\ &= \dots \\ &= S(\Delta \| \Delta_{\lambda}) + \text{constant?} \end{aligned}$$

Since $S(\Delta \| \Delta_{\lambda}) \leq 0$ with equality iff $\Delta = \Delta_{\lambda}$, we see that Δ_{λ} maximizes this. Equivalently,

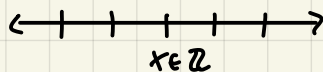
Gibbs states are the minimizers of F !

Note that β controls the relative weights of energy $\langle H_{\lambda} \rangle_{\Delta}$ and entropy $S(\Delta \| \mu)$ in this optimization. Temperature controls the balance between energy & entropy!

Phase Transitions

Consider an Ising model on \mathbb{Z}^d , where $\forall x \in \mathbb{Z}^d, \sigma_x \in \{-1, +1\}$, with an energy given by $H_{\lambda} = -\frac{J}{2} \sum_{\substack{u, v \\ \text{neighbors}}} \sigma_u \sigma_v$

1D Ising Model:



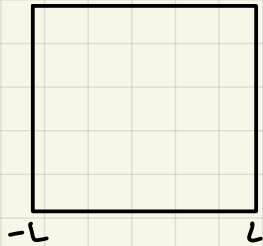
Note that as $T \rightarrow 0$, we only want to minimize H_{λ} , and so there are two ground states: +++++ and ----- (this is an example of **discrete symmetry breaking**, where a symmetry of the Hamiltonian (spin flip) leads to multiple distinct states).

We can show that for $T > 0$, there is no phase transition. We can manufacture a Markov chain where each flip is $\sim \text{Bernoulli}(p)$, and the length of that flip is $\sim \text{Exponential}(\mu)$: there is no phase transition in 1D!

← Even: go over this proof?

2D Ising Model:

We'd like to study the infinite limit. First, though, let's discuss things for a finite volume.



$$\Omega_0 = [-L, L]^2 \cap \mathbb{Z}^2$$

$$\sigma_x \in \{-1, +1\}$$

$$H = -\frac{J}{2} \sum_{u \sim v} \sigma_u \sigma_v$$

We expect majority to be the same sign, with some occasional clusters of flips.

Q: What would symmetry breaking look like?

A: We ask two questions:

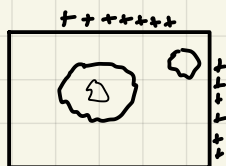
(1) if we apply external magnetic field $h \in \{\pm 1\}$, does the system $\mathcal{D}_\beta(h)$ have a discontinuity

\iff

(2) do the boundary conditions (+ or - along boundary of box) affect \mathcal{D}_β in the interior as $L \rightarrow \infty$?

We work with the second of these two formulations.

Def: A **Percus contour** is a closed path on \mathbb{Z}^2 s.t. the spins on its interior are the same, and are opposite the spins on the exterior.



Theorem: For the Ising model on \mathbb{Z}^2 , there \exists a β_c s.t. $\forall \beta > \beta_c$,

$$\mathbb{P}_{\beta, L}^{(+)} \left\{ \sigma_0 = -1 \right\} \leq p_0 \quad \forall L, \text{ where } p_0 < \frac{1}{2} \text{ doesn't depend on } L.$$

Proof: Let γ = a polygonal path w. $\sigma_x = \begin{cases} + & \text{on outside} \\ - & \text{on inside} \end{cases}$ be a Percus contour.

We claim that any arbitrary polygonal path has this property with probability $\leq e^{-2\beta J |\gamma|}$.

To see this, note that

$$\mathbb{P}_{\beta, L}^{(+)} \left\{ \gamma \text{ satisfies above} \right\} = \frac{\sum_{\sigma \in \mathcal{Z}_L} \mathbb{1}_\gamma \{ \sigma \} e^{-\beta H_L(\sigma)}}{\mathcal{Z}_L}$$

indicator for whether the state makes γ a Percus contour

configuration

let us use a rescaled and shifted Hamiltonian

$$H_L(\sigma) = 2J \sum_{\langle u,v \rangle} \left(\frac{1 - \sigma_u \sigma_v}{2} \right) + \text{const} = 2J \sum_{\gamma} \mathbb{1}_{\gamma} \{ \sigma \} + \text{const}$$

form a geometric Hamiltonian!

s.t. the energy counts the # of deviations from uniform +.

we can say $Z_L \geq \sum_{\sigma \in \Omega} \mathbb{1}_{\gamma} \{ \sigma \} \left[e^{-\beta H_L(\sigma)} + e^{-\beta H_L(R_{\gamma} \sigma)} \right]$
why?

where $(R_{\gamma} \sigma)_x = \begin{cases} -\sigma_x & \text{if } x \text{ is inside } \gamma \\ \sigma_x & \text{otherwise} \end{cases}$ is the mapping that flips spins inside γ .

$$\Rightarrow Z_L \geq \left(1 + e^{\beta H_L(\sigma) - \beta H_L(R_{\gamma} \sigma)} \right) \sum_{\sigma \in \Omega} \mathbb{1}_{\gamma} \{ \sigma \} e^{-\beta H_L(\sigma)}$$

Each contour satisfies $H_L(\sigma) - H_L(R_{\gamma} \sigma) = 2|\gamma|$ by our rewritten Hamiltonian.

$$\Rightarrow Z_L \geq \left(1 + e^{2\beta|\gamma|} \right) \sum_{\sigma \in \Omega} \mathbb{1}_{\gamma} \{ \sigma \} e^{-\beta H_L(\sigma)}$$

$$\Rightarrow \mathbb{P}_{\beta, L}^{(+)} \{ \gamma \text{ satisfies above} \} \leq \frac{1}{1 + e^{2\beta|\gamma|}}$$

the "energy estimate" of each γ ; they are unlikely, but there are many

We can now form the "entropy estimate", or the # of closed polygonal paths encircling the origin (or any interior point).

The number of γ 's encircling the origin s.t. $|\gamma| = m$ is $\frac{m}{2} 3^m$, *why?* and so by a union bound,

$$\begin{aligned} \mathbb{P}_{\beta, L}^{(+)} \{ \sigma_0 = -1 \} &= \mathbb{P}_{\beta, L}^{(+)} \{ \text{a contour encircles } x=0 \} \\ &\leq \sum_{\substack{\gamma \text{ encircling} \\ x=0}} \mathbb{P}_{\beta, L}^{(+)} \{ \gamma \text{ satisfies above} \} \leq \sum_{\substack{\gamma \text{ encircling} \\ 0}} e^{-2\beta|\gamma|} \\ &= \sum_{n=4}^{\infty} e^{-2\beta n} \cdot \frac{n}{2} 3^n = \sum_{n=4}^{\infty} \frac{n}{2} \cdot (3e^{-\beta})^n \end{aligned}$$

length must be ≥ 4 to encircle a point.

we finished this proof on a pset

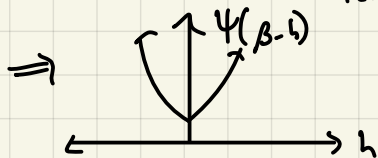
We just saw that when $\beta > \beta_c$,

$$\langle \sigma_x \rangle_{\lambda, \beta}^{(+)} \geq (1 - p_0) - p_0 = 1 - 2p_0 > 0 \quad \text{and} \quad \langle \sigma_x \rangle_{\lambda, \beta}^{(-)} = -\langle \sigma_x \rangle_{\lambda, \beta}^{(+)} < 0$$

Also, we had

magnetization
↓

$$m(\beta) = \frac{1}{|\Lambda|} \sum_{x \in \Lambda} \langle \sigma_x \rangle_{\lambda, \beta}^{(+)} = \beta \frac{\partial}{\partial h} \Psi(\beta, h)$$



Lecture 2/28 - Continuous Symmetry Breaking

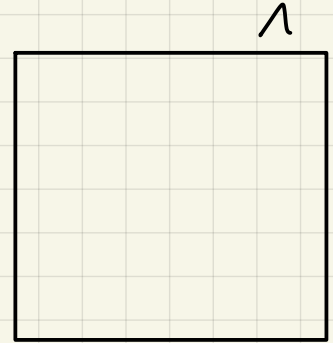
Note that Peierls argument of flip contains no longer works for vector-valued spins. Generalizing from systems with $\sigma_x \in \{-1, 1\}$ with global spin flip symmetry, we discuss N -dimensional Ising model with $O(N)$ symmetry and $\vec{\sigma}_x = (\sigma_{x,1}, \dots, \sigma_{x,N}) \in S^{N-1}$.

$O(N)$ -Symmetric Model

$$H_1^{(B.C.)} = - \sum_{(x,y) \in \Lambda^2} J_{x,y} \vec{\sigma}_x \cdot \vec{\sigma}_y - \sum_x \vec{h} \cdot \vec{\sigma}_x$$

$$= \frac{1}{2} \sum_{(x,y) \in \Lambda^2} J_{x,y} \|\vec{\sigma}_x - \vec{\sigma}_y\|_2^2 - \sum_x \vec{h} \cdot \vec{\sigma}_x$$

\uparrow
orthogonal N, N matrices \uparrow
unit N -sphere



Boundary Conditions

BC's of Λ can be:

- free

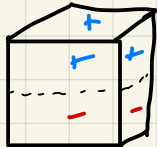
- uniform: $\sigma_x = (1, 0, \dots) \forall x \in \partial \Lambda$

- periodic: In 1D, $H = - \sum_{n=1}^{L-1} \vec{\sigma}_n \cdot \vec{\sigma}_{n+1} - \vec{\sigma}_L \cdot \vec{\sigma}_1$

this corresponds to gluing the chain into a ring!

These all generate translation-invariant states in the thermodynamic limit!

In fact, all B.C.'s in 2D yield translation-invariant states. In 3D, we can construct



We select periodic B.C.'s.

Fourier Transform

For $\Lambda_L = [-\frac{L}{2}, \frac{L}{2}]^d$, let $\Lambda_L^* = (-\pi, \pi]^d \cap \frac{\pi}{L} \mathbb{Z}^d$.

We have the transform and its inverse given by

$$\hat{\sigma}(\vec{p}) = \frac{1}{\sqrt{|\Lambda_L|}} \sum_{\vec{x} \in \Lambda_L} e^{-i\vec{p} \cdot \vec{x}} \vec{\sigma}_x$$

$$\vec{\sigma}_x = \frac{1}{\sqrt{|\Lambda_L|}} \sum_{\vec{p} \in \Lambda_L^*} e^{i\vec{p} \cdot \vec{x}} \hat{\sigma}(\vec{p})$$

All arbitrary spin configurations can be seen as superpositions of plane waves!

We can verify that they are inverses.

$$\text{RHS} = \frac{1}{\sqrt{|\Lambda_L|}} \sum_{\vec{p} \in \Lambda_L^*} e^{i\vec{p} \cdot \vec{x}} \frac{1}{\sqrt{|\Lambda_L|}} \sum_{\vec{u} \in \Lambda_L} e^{-i\vec{p} \cdot \vec{u}} \vec{\sigma}_u = \sum_{\vec{u} \in \Lambda_L} \vec{\sigma}_u \frac{1}{|\Lambda_L|} \sum_{\vec{p} \in \Lambda_L^*} e^{i\vec{p} \cdot (\vec{x} - \vec{u})} = \sum_{\vec{u} \in \Lambda_L} \vec{\sigma}_u \delta_{\vec{x}-\vec{u}} = \vec{\sigma}_x$$

$= \delta_{\vec{x}-\vec{u}}$ since $\sum_{\vec{p} \in \Lambda_L^*} e^{i\vec{p} \cdot (\vec{x}-\vec{u})} = \sum_{\vec{p} \in \Lambda_L^*} e^{i\vec{p} \cdot (\vec{x}-\vec{u})} \approx \int_{\text{by steepest desc}} e^{i\vec{p} \cdot (\vec{x}-\vec{u})} d\vec{p} = 0$

Suppose that $\vec{h}=0$ (no external field).

Now, we want to write H in terms of $\hat{\sigma}(\vec{p})$. Note that $\vec{h}=0 \Rightarrow H_L = (\sigma, A\sigma) = \sum \hat{\sigma}_y^T J_{y,x} \hat{\sigma}_x$ is simply a matrix product. Since it is translation-invariant, we know that it is simultaneously diagonalizable with the translation operator. These are the plane waves, i.e. $\psi_{\vec{p}}(\vec{x}) = \frac{1}{\sqrt{N_L}} e^{i\vec{p}\cdot\vec{x}} \Rightarrow \psi_{\vec{p}}(\vec{x}+\vec{a}) = e^{i\vec{p}\cdot\vec{a}} \psi_{\vec{p}}(\vec{x})$

Consider the function $\Psi(x) = \hat{\sigma}_x \Rightarrow |\Psi\rangle = \sum_{p \in \Lambda_L^*} \langle \psi_p | \Psi \rangle |\psi_p\rangle$ (project to orthonormal basis $\{|\psi_p\rangle\}_p$ of L^2)

Then, $H = \langle \Psi | \hat{H} | \Psi \rangle = -\frac{1}{2} \langle \Psi | J | \Psi \rangle = -\frac{1}{2} \sum_{p,p'} \langle \Psi | \psi_p \rangle \langle \psi_p | J | \psi_{p'} \rangle \langle \psi_{p'} | \Psi \rangle = \sum_p |\langle \Psi | \psi_p \rangle|^2 \cdot \frac{1}{2} \langle J \rangle_{\psi_p}$

Equivalently, if we write H in diagonal form, we get

$$H = -\frac{1}{2} \sum_{x,y \in \Lambda_L} \hat{\sigma}_x^T J_{x,y} \hat{\sigma}_y$$

$$= -\frac{1}{2} (\sigma, J \sigma)$$

$J_{x,y} = \begin{cases} 1 & \|x-y\|=1 \\ -2d & \|x-y\|=0 \\ 0 & \text{otherwise} \end{cases}$

J is basically a discrete version of the Laplacian (discrete differences)

We have $J \psi_{\vec{p}}(\vec{x}) = \frac{1}{\sqrt{N_L}} e^{i\vec{p}\cdot\vec{x}} \sum_{\vec{u}} (1 - e^{i\vec{p}\cdot\vec{u}}) J_{\vec{x},\vec{x}+\vec{u}}$

Computing,

$$\xi(\vec{p}) = \frac{1}{2} \sum_{\|\vec{u}\|=1} (1 - e^{i\vec{p}\cdot\vec{u}}) = -\frac{1}{2} \sum_{j=1}^d 2(1 - \cos(p_j)) = 2 \sum_{j=1}^d \sin^2\left(\frac{p_j}{2}\right)$$

For small p_j 's, this behaves like $\xi(\vec{p}) \approx \frac{1}{2} \sum_j p_j^2 = \frac{1}{2} \|\vec{p}\|^2$ (kinetic energy is $\frac{p^2}{2}$!)

slaky stuff. his "nose was too close to the blackboard"

The energy is the sum of the energies of the plane waves, yielding

$$H_L = \sum_{p \in \Lambda_L^*} \xi(\vec{p}) |\hat{\sigma}(\vec{p})|^2 - \frac{1}{2} \sum_{p \neq 0} \xi(\vec{p}) |\hat{\sigma}(\vec{p})|^2$$

with $\xi(\vec{p}) = \sum_{u \in \Lambda_L} e^{i\vec{p}\cdot\vec{u}} J_{\vec{u}}$ (F.T. of coupling energy)

This agrees with our bracket stuff. In total, we get that the energy decomposes to the sum of plane wave energies! We also know that if $\xi(\vec{p}) = \frac{p^2}{2m}$, $\langle \xi(\vec{p}) \rangle = \frac{1}{2} k_B T$

Now, Parseval (f and \hat{f} have same L^2 norm) yields

$$1 = \frac{1}{N_L} \sum_{x \in \Lambda_L} |\sigma_x|^2 = \frac{1}{N_L} \sum_{p \in \Lambda_L^*} |\hat{\sigma}(\vec{p})|^2$$

$$\Rightarrow 1 = \frac{1}{N_L} \sum_{p \in \Lambda_L^*} \langle |\hat{\sigma}(\vec{p})|^2 \rangle \approx \frac{1}{2} \frac{1}{N_L} \sum_{p \neq 0} \frac{1}{|p|^2} \langle \|\hat{\sigma}(\vec{p})\|^2 \xi(\vec{p}) \rangle + \frac{1}{2N_L} \langle \|\hat{\sigma}(\vec{0})\|^2 \rangle$$

$\sim \int_{(-\pi, \pi)^d} \frac{1}{|p|^2}$, diverges if $d \leq 2$

Lastly, note that the Fourier transform of spin-spin correlation functions appears as

$$\hat{S}_{\Delta}^{(L)}(\vec{p}) := \sum_{x \in \Lambda_L} e^{i\vec{p}\cdot\vec{x}} S_{\Delta}^{(L)}(\vec{x}) = \sum_{x \in \Lambda_L} e^{i\vec{p}\cdot\vec{x}} \langle \hat{\sigma}_0 \cdot \hat{\sigma}_x \rangle_{L_c} = \langle \|\hat{\sigma}(\vec{p})\|^2 \rangle_{L_c}$$

intensity of the p th mode

Symmetry Breaking as a Condensation Phenomenon

The above reasoning, together with the **equipartition law**, allow us to give a sufficient condition for symmetry breaking which leads to condensation into macroscopic occupation of the ground state (a la Bose-Einstein Condensation).

Prop 8.1: needed for local integrability of $\frac{1}{|\hat{p}|^2}$

Let $d \geq 2$. Suppose that in a system of bounded spins with nearest-neighbor interaction, we have the **Gaussian domination bound**

$$\mathcal{E}(\rho) \hat{S}_{\rho, \beta}^{(L)}(\rho) \leq \frac{1}{2\beta} \quad \forall \rho$$

Define

$$C_d := \frac{1}{(2\pi)^d} \int_{[-\pi, \pi]^d} \frac{1}{\mathcal{E}(\hat{p})} d\hat{p}$$

Then, $\forall \beta > C_d/2$, the following hold

- (i) $\liminf_{L \rightarrow \infty} \left\langle \left\| \frac{1}{|A_L|} \sum_{x \in A_L} \sigma_x \right\|^2 \right\rangle \geq 1 - \frac{C_d}{2\beta}$ (expected magnitude of bulk magnetization increases with β)
- (ii) at $\tilde{h}=0$, $\Psi(\beta, \tilde{h})$ has discontinuous derivative (core singularity) (phase transition!)
- (iii) in the infinite limit, the system has Gibbs states of nonzero magnetization, i.e. the spin-rotation symmetry is broken

Proof: (i) Parseval-Plancherel yields that when spins are unit magnitude ($\|\hat{\sigma}(x)\|=1 \quad \forall x$),

$$\frac{1}{|A_L|} \|\hat{\sigma}(0)\|^2 + \frac{1}{|A_L|} \sum_{\rho \in \Lambda_L^* \setminus \{0\}} \|\hat{\sigma}(\rho)\|^2 = 1$$

$$= \left\| \frac{1}{|A_L|} \sum_{x \in A_L} \hat{\sigma}(x) \right\|^2$$

just the norm of $\hat{f}(0)$ is avg. value of f

Take an expectation,

$$\left\langle \left\| \frac{1}{|A_L|} \sum_{x \in A_L} \hat{\sigma}(x) \right\|^2 \right\rangle = 1 - \frac{1}{|A_L|} \sum_{\rho \in \Lambda_L^* \setminus \{0\}} \langle \|\hat{\sigma}(\rho)\|^2 \rangle = 1 - \frac{1}{|A_L|} \sum_{\rho \in \Lambda_L^* \setminus \{0\}} \hat{S}_{\rho, \beta}^{(L)}(\rho)$$

The Gaussian domination bound leads

$$\geq 1 - \frac{1}{2\beta} \cdot \left[\frac{1}{|A_L|} \sum_{\rho \in \Lambda_L^* \setminus \{0\}} \frac{1}{\mathcal{E}(\rho)} \right] \xrightarrow{L \rightarrow \infty} 1 - \frac{1}{2\beta} C_d$$

Riemann approx for C_d

(ii) From (i), we know that $\exists B > 0$ s.t. $\forall L$ large enough and all b.c.'s,

$$\left\langle \left\| \frac{1}{|A_L|} \sum_{x \in A_L} \vec{\sigma}(x) \right\|^2 \right\rangle_{\Lambda_L, \beta, \vec{h}=0}^{(b.c.)} \geq B^2$$

The finite-volume pressure function satisfies (with $\vec{h} = (1, 0, \dots)$)

$$e^{(\Psi(\beta, \vec{h}) - \Psi(\beta, 0)) |A_L|} = \left\langle e^{\beta \vec{h} \cdot \sum_{x \in A_L} \vec{\sigma}_x} \right\rangle \geq e^{\beta \|\vec{h}\| |A_L| B(1-\epsilon)} \mathbb{P} \left\{ \sum_{x \in A_L} \sigma_x^{(1)} \geq B(1-\epsilon) \right\}$$

A Chebyshev-type estimate gives

$$B^2 \leq \left\langle \left| \frac{1}{|A_L|} \sum_{x \in A_L} \sigma_x^{(1)} \right|^2 \right\rangle_{\vec{h}=0} \leq B^2(1-\epsilon)^2 + B^2 \mathbb{P} \left\{ \sum_{x \in A_L} \sigma_x^{(1)} \geq B(1-\epsilon) \right\}$$

$$\Rightarrow e^{(\Psi(\beta, \vec{h}) - \Psi(\beta, 0)) |A_L|} \geq e^{\beta \|\vec{h}\| |A_L| B(1-\epsilon)} \cdot \frac{B^2(1-(1-\epsilon)^2)}{B^2}$$

$$\stackrel{?}{\Rightarrow} \Psi(\beta, \vec{h}) - \Psi(\beta, 0) \geq \beta B \|\vec{h}\| \quad (\text{Lemma 8.2 in notes})$$

In particular, this implies a conical singularity at $\vec{h}=0$.

(iii) As always, discontinuous derivative of $\Psi \Rightarrow$ symmetry breaking. More explicitly, we have the relation

$$\left\langle \frac{1}{|A_L|} \sum_{x \in A_L} \vec{\sigma}(x) \right\rangle = \frac{1}{\beta} \vec{\nabla}_{\vec{h}} \Psi(\beta, \vec{h})$$

When $\vec{\nabla}_{\vec{h}} \Psi$ has different directional derivatives (singularity), to each direction corresponds at least one translation-invariant Gibbs state μ for which $\frac{1}{|A_L|} \sum_{x \in A_L} \vec{\sigma}(x) \stackrel{\mu\text{-a.s.}}{=} \text{the value of } \vec{\nabla}_{\vec{h}} \Psi \text{ in this direction}$

Each such μ exhibits rotational symmetry breaking. □

The condition $\mathcal{E}(\rho) \hat{S}_{\rho, \beta}^{(L)}(\rho) \leq \frac{1}{2\beta}$ is in general

not well-understood, and we only know it holds for

reflection-positive systems.

Lecture 3/7-

"What's a factor of 2 among friends?"

Remarks on Symmetry Breaking

If a system's Hamiltonian H and a priori distribution $\mu(d\sigma)$ are invariant under a certain transformation $\sigma_x \mapsto R(\sigma_x)$, this is a **symmetry**.

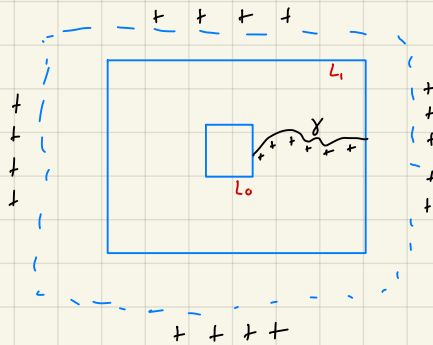
We say that a symmetry is **broken** if there exist β , an observable F , and a pair of boundary conditions bc_1, bc_2 such that

$$\langle F(\sigma) \rangle_{\beta}^{bc_1} = \lim_{L \rightarrow \infty} \langle F(\sigma) \rangle_{\Lambda_L, \beta}^{bc_1} \neq \lim_{L \rightarrow \infty} \langle F(R\sigma) \rangle_{\Lambda_L, \beta}^{bc_2} := \langle F(R\sigma) \rangle_{\beta}^{bc_2}$$

F is asymmetric under R in large volume limit

With $F(\sigma) = \sigma_0$, this translates to asking whether an interior point **remembers** for any **boundary conditions**.

Note: The boundary conditions are very far away, and observables we construct to prove symmetry breaking (Peierls argument) is



$$|c| \ll L_0 \ll L_1$$

$$F(\sigma) = \mathbb{1} \left\{ \exists \text{ a path with } + \text{ spins connecting } L_0 \text{ and } L_1 \right\}$$

$$F(\sigma) = \frac{1}{|\Lambda_{L_1}|} \sum_{x \in \Lambda_{L_1}} \tilde{\sigma}(x) \text{ for continuous case}$$

Back to Continuous Symmetry Breaking

Def: A vector space \mathcal{H} (over \mathbb{C}) is a **Hilbert space** if it has a positive inner product $\langle \cdot, \cdot \rangle$ s.t. $\forall f, g, h \in \mathcal{H}$,

$$(i) \langle f, g \rangle = \overline{\langle g, f \rangle}$$

$$(ii) \langle h, f+ag \rangle = \langle h, f \rangle + a \langle h, g \rangle$$

$$(iii) \langle f, f \rangle \geq 0$$

Theorem: (Schwartz Inequality)

$$\forall f, g \in \mathcal{H},$$

$$\langle f, g \rangle \leq \langle f, f \rangle^{1/2} \cdot \langle g, g \rangle^{1/2}$$

Proof: $\forall \lambda, \langle \lambda f, \lambda f \rangle \geq 0 \Rightarrow |\lambda|^2 \underbrace{\langle f, f \rangle}_{A^2} + \underbrace{\lambda \langle f, g \rangle + \bar{\lambda} \langle g, f \rangle}_{= 2 \operatorname{Re}(\lambda \langle f, g \rangle)} + \underbrace{\langle g, g \rangle}_{B^2} \geq 0$

$\Rightarrow \forall \lambda, A^2 (|\lambda|^2 + 2 \lambda \frac{C}{A^2}) + B^2 \geq 0 \Rightarrow A^2 (\lambda + \frac{C}{A^2})^2 - A^2 \frac{C^2}{A^4} + B^2 \geq 0$

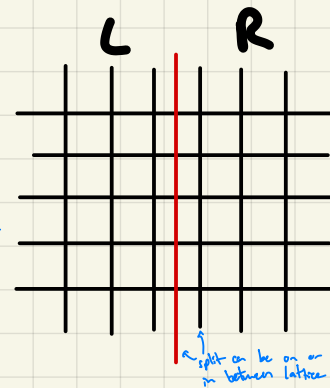
Letting $\lambda = -\frac{C}{A^2}, -\frac{C^2}{A^2} + B^2 \geq 0 \Rightarrow |C| \leq \sqrt{AB}$

□

Def: Let $\mathcal{H}_+ = \{f: f \in \mathcal{B}_+ \text{ w. } \mathbb{E}[f^2] < \infty\}$

be a Hilbert space of measurable functions that depend only on states on the right side of a split, with inner product

$$\langle F, G \rangle = \int \bar{F}(\sigma) \mathcal{R} G(\sigma) \mu_{\mathcal{R}}(d\sigma) = \mathbb{E}[\bar{F} \mathcal{R} G]$$



We say a system has **reflection positivity** about a reflection \mathcal{R} iff $\forall F, G: \mathcal{R} \rightarrow \mathbb{C}$ observables in $\mathcal{B}_+,$

$$\langle F, F \rangle = \mathbb{E}[\bar{F} \mathcal{R} F] \geq 0 \quad \text{and} \quad \langle F, G \rangle = \overline{\langle G, F \rangle}$$

To characterize which systems have RP, a sufficient condition is

Prop 8.4:

A sufficient condition for the Gibbs states to be RP w.r.t. a reflection \mathcal{R} is that its Hamiltonian can be written as

$$-H = A + \mathcal{R}A + \sum_{j=1}^k B_j \mathcal{R}B_j \quad \text{where } A, B_j \in \mathcal{C}_+ \text{ depend only on spins on one side of } \mathcal{R}$$

Proof: check the notes \therefore

□

Some other conditions for reflection positivity and examples of long-range RP interactions are presented in Fradki/Velenik and Fröhlich/Zegarliński:

For $1 \leq d \leq 4,$ this class includes two-body spin-spin interactions with power-law decay like $\int_{x-y} = \frac{1}{\|x-y\|^\gamma}, \quad \gamma \geq |d-2|_+$

The Chessboard Inequality

Note first that Schwartz + RP gives that $\forall F, G \in \mathcal{H}_+$,

$$\mathbb{E}[\overline{F}RG] \leq \mathbb{E}[\overline{F}RF]^{\frac{1}{2}} \mathbb{E}[\overline{G}RG]^{\frac{1}{2}}$$

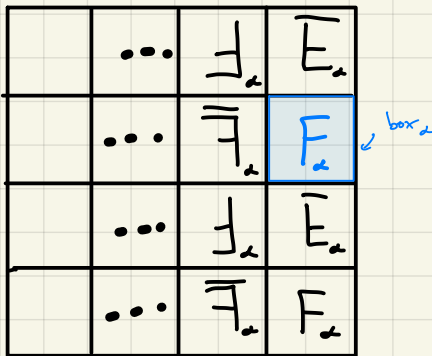
Consider a reflection R , and let \mathcal{B}_{\pm} be the collection of functions depending only on spins in Λ_{\pm} ; i.e. \mathcal{B}_{\pm} are functions measurable w.r.t. $\mathcal{O}(\Omega_{\Lambda_{\pm}})$.

Then, $F_+ \in \mathcal{B}_+$ and $F_- \in \mathcal{B}_-$ gives through CS+RP that, since $F_- = RG$ for some $G \in \mathcal{B}_+$,

$$\begin{aligned} |\mathbb{E}[F_+ F_-]|^2 &= |\mathbb{E}[F_+ RG]|^2 \leq \mathbb{E}[F_+ \overline{R}F_+] \mathbb{E}[\overline{G}RG] \\ &= \mathbb{E}[F_+ R F_+] \mathbb{E}[\overline{R}F_- F_-] = \mathbb{E}[F_+ R F_+] \mathbb{E}[\overline{F}_- R F_-] \end{aligned}$$

Suggestively written, this means that the expectation of a product $F_+(\omega_+) F_-(\omega_-)$ is bounded by the geometric mean of $\mathbb{E}[\overline{F}_{\pm} R F_{\pm}]$, done by reflecting and conjugating throughout both domains $\Omega_{\Lambda_{\pm}}$. Generalizing to more domains,

Theorem (Chessboard Inequality)



If a spin system in Λ is RP w.r.t. a family of reflections across perpendicular hyperplanes that divide Λ into almost-disjoint boxes $\Lambda = \bigcup_{\alpha} \Lambda_{\alpha}$, then \forall families $\{F_{\alpha}\}$ with $F_{\alpha} \in \mathcal{B}_{\Lambda_{\alpha}}$, only depend on spins in box Λ_{α}

$$\left| \mathbb{E} \left[\prod_{\alpha} F_{\alpha}(\omega_{\alpha}) \right] \right| \leq \prod_{\alpha} \mathbb{E} \left[\prod_{\gamma} F_{\alpha}^{\#}(\omega_{\gamma}) \right]^{\frac{1}{\# \text{ boxes}}} \geq 0 \text{ by RP}$$

"The product is dominated by what happens if you take them one at a time and duplicate them everywhere."

Proof: WOLOG (by scaling), suppose that $\mathbb{E} \left[\prod_{\gamma} F_{\alpha}^{\#}(\omega_{\gamma}) \right] = 1 \quad \forall \alpha$.

This reduces the task to proving the following statement:

Let $\mathcal{S} = \{F_j\}_{j=1, \dots, K} \subset \mathcal{B}_{\Lambda_0}$ be a collection of functions measurable in a common box Λ_0 , each normalized by (8.38), and let $\kappa : \{1, \dots, K\} \rightarrow \{1, \dots, K\}$ represent assignments of functions from \mathcal{S} to the cells. Then the following maximum

$$\max_{\kappa: \{1, \dots, K\} \rightarrow \{1, \dots, K\}} \left| \left\langle \prod_{\alpha} F_{\kappa(\alpha)}^{\#}(\sigma_{\alpha}) \right\rangle \right| \quad (8.39)$$

(which need not be unique) is attained by a configuration for which $\kappa(\alpha)$ is constant.

By the Cauchy-Schwarz inequality if κ is maximizer then so is each of the two configurations which are obtained by symmetrizing κ with respect to an arbitrarily chosen reflection plane. Such reflections can be used to decrease the amount of disagreement in the nearest neighbor assignments while staying within the collection of optimizing assignments. The only maximizing configurations whose nearest neighbor disagreement cannot be further reduced corresponds to κ such that $\kappa(\alpha) = \kappa(\alpha')$ for each pair of neighboring boxes. This condition implies that among the maximizer there is one for which $\kappa(\alpha)$ takes a common value for all α , and the claim follows. \square

?? huh??

"Feed FSS and FLS for more"

The Gaussian Domination Bound

Working once again in an $O(N)$ spin model with

$$H_n^{\text{per}}(\sigma) = \frac{1}{2} \sum_{\substack{x, y \\ |x-y| \leq 1}} J_{xy} \|\sigma_x - \sigma_y\|^2 \quad \left(-h \cdot \sum_{x \in \Lambda} \vec{\sigma}_x \right)$$

we suppose further that corresponding Gibbs states are RP (this is automatically the case) for such models

Recall our partition function

$$Z_{\Lambda} = \int_{\sigma \in \mathcal{R}_{\Lambda}} e^{-\frac{\beta}{2} \sum_{x, y} J_{xy} (\sigma_x - \sigma_y)^2}$$

Consider a modified partition function

$$Z_{\Lambda}(\vec{z}) = \int_{\sigma \in \mathcal{R}_{\Lambda}} e^{-\frac{\beta}{2} \sum_{x, y} J_{xy} (\sigma_x + z_x - \sigma_y + z_y)^2}$$

Here, \vec{z} denotes a stress/biasing of the spins at each site.

Theorem:

$$\forall \vec{z}, \quad Z_{\Lambda}(\vec{z}) \leq Z_{\Lambda}$$

trivial measure!

Proof:
$$\frac{Z_{\Lambda}(\vec{z})}{Z_{\Lambda}} = \int_{\mathcal{R}^{|\Lambda|}} e^{-\frac{\beta}{2} \sum (\sigma_x - \sigma_y)^2} \frac{\prod_x (T_{z_x} \mu_0)(d\sigma_x)}{Z_{\Lambda}}$$

$$= \mathbb{E} \left[\prod_{x \in \Lambda} \delta \right]$$

finish proof that $O(N)$ models over \mathbb{Z}^d with interactions of the n.n. form (which are RP) satisfy

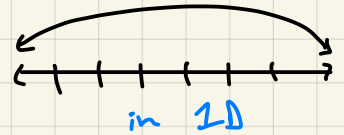
$$\hat{S}_{A, B}^{(1)}(\vec{p}) \mathcal{E}(\vec{p}) \leq \frac{1}{2\beta}$$

Lecture ??? - Transfer Matrices

Finish \rightarrow

Correlators for 1D

"as the Chinese proverb says, there are a dozen ways to cook rice"



Consider a system of spins with periodic b.c.'s :

Writing $T_{\sigma\sigma}$ as a transfer matrix and diagonalizing

$$T = \begin{pmatrix} \lambda_1 & & \dots & \\ & \ddots & & \\ & & \lambda_k & \\ & & & \ddots \end{pmatrix} \Rightarrow T^L = \begin{pmatrix} \lambda_1^L & & \dots & \\ & \ddots & & \\ & & \lambda_k^L & \\ & & & \ddots \end{pmatrix} \quad \text{let } \lambda_1, \lambda_2 \quad \forall j \geq 2$$

and so for $k=2$ with n.n. interactions,

$$Z_L^{\text{per}} = \sum_{\sigma_1, \dots, \sigma_n} = \text{tr}(T^L) = \sum_{i=1}^k \lambda_i^L \stackrel{k=2}{=} \lambda_1^L (1 + e^{-\alpha L})$$

and

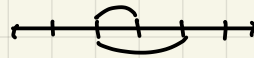
$$\langle \sigma_x; \sigma_y \rangle_L^{\text{per}} := \langle \sigma_x \sigma_y \rangle_L^{\text{per}} - \langle \sigma_x \rangle_L^{\text{per}} \langle \sigma_y \rangle_L^{\text{per}} = |\langle \psi_1 | S | \psi_2 \rangle|^2 e^{-\alpha d(x,y)} \left[1 + O(e^{-\frac{\alpha L}{2}}) \right]$$

for periodic b.c.'s, $d(x,y) = \min\{|x-y|, L-|x-y|\}$

connected correlation decays exponentially fast!

truncated correlator

If we no longer have n.n. interactions, such as
In this case, we define $\chi_n := (\sigma_n^{(1)}, \sigma_n^{(2)})$ and the transfer matrix



$$T_{n,n+1} = \prod [\chi_n^{(2)}, \chi_{n+1}^{(1)}] e^{-\beta V_{n,n+1}} \Rightarrow T \in \mathbb{R}^{4 \times 4} \Rightarrow T^2 \text{ is positive?}$$

(ensures the overlapping indices make sense)

Alternatively, we can group each two sites into a single site and write a more complex $T \in \mathbb{R}^{2 \times 2}$.

Lecture 3/21 - Infinite-Volume Gibbs States

(If you didn't get the right answer for 4.2, you can do it again and email it to the grader and CC Arzuman.)

Note that upon taking a limit, we both gain and lose information. We may lose boundary conditions, and we may gain translation invariance, etc. So, it makes sense to consider **Gibbs states** in the infinite-volume limit.

Recall: For finite volumes Λ , Gibbs states form probability measures on Ω_Λ with density

$$\Delta(dw) = \frac{e^{-\beta H_\Lambda(w)}}{Z_\Lambda} \mu_\Lambda(dw)$$

\uparrow
configuration space
 $\xrightarrow{\text{the infinite limit of this density makes sense}}$
 $\mu_\Lambda(dw)$
a-priori measure

In the Ising model, $\Omega_\Lambda = \{-1, 1\}^\Lambda \Rightarrow w \in \Omega_\Lambda$ is a map $w: \Lambda \rightarrow \{-1, 1\}$

In the infinite limit: For the Ising model, $\Omega = \{-1, 1\}^{\mathbb{Z}^d}$, where $w \in \Omega$ is a map $w: \mathbb{Z}^d \rightarrow \{-1, 1\}$ and $\sigma_x = w(x)$.

Note that this infinite sequence of binary choices is exactly like how we describe $[0, 1]$ via binary expansion. So, $\Omega \cong [0, 1]$. Now, let us investigate the topology of Ω .

First, we will need a crash course in some measure theory and conditional probability.

Some measure theory

In the σ -algebra of measurable sets, we must certainly have all **local sets** - sets which are describable by a local characterization.

examples

- collections of configurations $\{w \mid \exists \begin{matrix} \boxed{\text{---}} \\ \text{---} \end{matrix} \}$ \leftarrow local description

- any set for which inclusion can be verified by looking at a finite region

So, we can define \mathcal{B} \leftarrow measurable sets to be the minimal σ -algebra containing the local sets.

Def: A function $f: \Omega \rightarrow \mathbb{R}$ is **measurable** w.r.t. a σ -algebra \mathcal{B} if and only if $\{w \in \Omega : f(w) < \alpha\} \in \mathcal{B}$ for all $\alpha \in \mathbb{R}$ (preimages of $f < \alpha$ are measurable)

We can ask the following question:

$$\text{is } f(\omega) = \limsup_{L \rightarrow \infty} \frac{1}{|\Lambda_L|} \sum_{x \in \Lambda_L} \sigma_x \text{ measurable?}$$

↑
sequence
(σ_x)_x

Note that this function is very nonlocal: no finite set $A \subseteq \Lambda$ determines the value of f . However, we can show that f is measurable!

The condition $\{f(\omega) < 2\}$ is equivalent to the event that $\forall \epsilon = \frac{1}{k} > 0, \exists N(k)$ s.t. $\forall L > N(k), \frac{1}{|\Lambda_L|} \sum_{x \in \Lambda_L} \sigma_x < 2 + \frac{1}{k}$

this is a local condition!
 $\Rightarrow \in \mathcal{B}$

Let $A_{L,k}$ be the set $A_{L,k} := \left\{ \omega \in \Omega : \frac{1}{|\Lambda_L|} \sum_{x \in \Lambda_L} \sigma_x < 2 + \frac{1}{k} \right\} \in \mathcal{B}$

Then, we can write $A_2 := \left\{ \omega \in \Omega : f(\omega) < 2 \right\} = \bigcap_{k=1}^{\infty} \bigcap_{L=N(k)}^{\infty} A_{L,k}$

By closure of \mathcal{B} under countable intersection, $A_2 \in \mathcal{B} \Rightarrow f$ is measurable.

Lastly, let us define $\mathcal{B}_{\infty} := \mathcal{B} \bigcap_{\substack{\Lambda \subseteq \mathbb{Z}^d \\ |\Lambda| < \infty}} (\mathcal{B}_{\Lambda})^c$. In words, \mathcal{B}_{∞} denotes

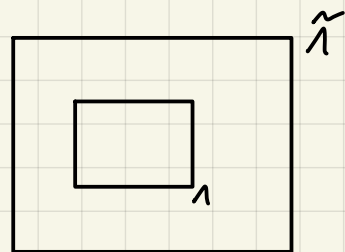
events measurable within Λ

the measurable sets that don't depend on any finite region.

Some conditional probability

Consider two finite volumes $\Lambda \subseteq \tilde{\Lambda}$, and define $\Lambda^c := \tilde{\Lambda} \setminus \Lambda$.

Suppose that we'd like to know the distribution of configurations in Λ given the configuration in Λ^c .



Note that $H(\omega_{\Lambda}, \omega_{\Lambda^c}) = H_{\Lambda}(\omega_{\Lambda}; \omega_{\Lambda^c}) + H_{\Lambda^c}(\omega_{\Lambda^c})$

$H_{\Lambda}(\omega_{\Lambda}; \omega_{\Lambda^c})$ contains things inside Λ and interactions between Λ and Λ^c ; in a sense, ω_{Λ^c} determines the boundary conditions. We can write out the **conditional Gibbs measure**

$$\mu(d\omega_{\Lambda} | \omega_{\Lambda^c}) = \frac{e^{-\beta H_{\Lambda}(\omega_{\Lambda}; \omega_{\Lambda^c})}}{Z_{\Lambda; \omega_{\Lambda^c}}} \mu(d\omega_{\Lambda})$$

The above expression led DLR to define the infinite Gibbs measure for $|\tilde{\Lambda}| \rightarrow \infty$ as:

Def: An infinite Gibbs state for a Hamiltonian

$$H(\omega) = \sum_{A \subset \mathbb{Z}^d} \int_A \phi_A(\omega_A)$$

is any probability measure ν on (Ω, \mathcal{B}) whose finite volume conditional probability is

$$\nu(dw_\Lambda | w_{\Lambda^c}) = \frac{e^{-\beta H_\Lambda(\omega_\Lambda; w_{\Lambda^c})}}{Z_{\Lambda; w_{\Lambda^c}}} \mu(dw_\Lambda)$$

This formulation gives a good characterization of symmetry breaking! We say that there is symmetry breaking if there are infinite Gibbs states whose densities don't have symmetries that the system (H, μ) have.

Lecture 3/23-

Regular Conditional Expectation

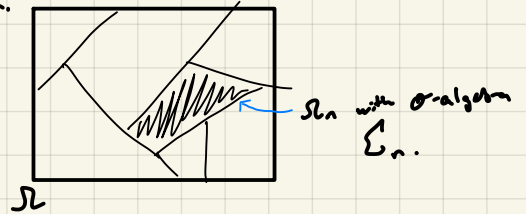
Recall from probability theory the following discussion on **regular conditional expectation**:

In the beginning, we had $\mathbb{P}\{A|B\} := \frac{\mathbb{P}\{A \cap B\}}{\mathbb{P}\{B\}}$

More generally, consider a probability space Ω partitioned into finite $(\Omega_n)_n$ that are disjoint. For each n , define Σ_n as its σ -algebra.

Then, we define

$$\mathbb{E}[f | \Sigma_n](\omega) := \frac{\int_{\Omega_n} f(\omega) \mu(d\omega)}{\int_{\Omega_n} \mu(d\omega)}$$



From here, we generalize to general σ -algebras. More formally, we have the existence of **regular conditional expectation**

Prop 10.7:

Let (Ω, Σ, μ) be a probability space and $\Sigma_0 \subseteq \Sigma$ a sub- σ -algebra. Then, \exists a unique linear map associating to each bounded, Σ -measurable function $f \in L^\infty(\Omega, \Sigma)$ the function $\mathbb{E}[f | \Sigma_0]: \Omega \rightarrow \mathbb{C}$ s.t.

(i) $\mathbb{E}[f | \Sigma_0] \in L^\infty$

(ii) $\forall f \in L^\infty(\Omega, \Sigma)$ and all $g \in L^\infty(\Omega, \Sigma_0)$, Σ_0 -measurable

$$\int_{\Omega} f(\omega) g(\omega) \mu(d\omega) = \int_{\Omega} \mathbb{E}[f | \Sigma_0](\omega) g(\omega) \mu(d\omega)$$

Remarks:

① In the L^2 perspective, the mapping $P_{\Sigma_0}: L^2(\Omega, \Sigma) \rightarrow L^2(\Omega, \Sigma)$ extends in $L^2(d\mu)$ into an orthogonal projection $f \mapsto \mathbb{E}[f | \Sigma_0]$ onto the subspace

$$\text{Range } P_{\Sigma_0} = \{f \in L^2(d\mu) : f \text{ is } \Sigma_0\text{-measurable}\}$$

② \forall monotone decreasing sequences of σ -algebras $\Sigma_1 \supseteq \dots \supseteq \Sigma_n \supseteq \dots$, the corresponding projections commute and have the tower property

$$P_{\Sigma_n} P_{\Sigma_k} = P_{\Sigma_n} \quad \forall n \geq k \quad \text{i.e.,} \quad \mathbb{E}[\mathbb{E}[f | \Sigma_k] | \Sigma_n] = \mathbb{E}[f | \Sigma_n] \quad \text{for } \Sigma_n \subseteq \Sigma_k$$

In probabilistic terms, for bounded f , $\{P_{\Sigma_n} f\}_n$ forms a martingale.

③ By the martingale convergence theorem, $\forall f \in L^\infty(\Omega, \Sigma)$ the pointwise limit $\lim_{n \rightarrow \infty} P_{\Sigma_n} f(\omega)$ exists μ -a.s. and yields the function $P_{\Sigma_\infty} f$, $\Sigma_\infty = \bigcap_n \Sigma_n$.

We have the following theorem:

Theorem: (Dobrushin-Lanford-Ruelle Condition)

For all finite $\Lambda \subseteq \mathbb{Z}^d$,

$$\mathbb{E}[f(\sigma_\Lambda, \sigma_{\Lambda^c}) | \sigma_{\Lambda^c}] = \int_{\Omega_\Lambda} f(\sigma_\Lambda, \sigma_{\Lambda^c}) \frac{e^{-\beta H_\Lambda(\sigma_\Lambda | \sigma_{\Lambda^c})}}{Z_\Lambda} \mu_\Lambda(d\sigma_\Lambda)$$

In the probability theory notation,

$$\mathbb{E}[f | \Sigma_{\Lambda^c}](\omega) = \int_{\Omega_\Lambda} f(\sigma_\Lambda, \sigma_{\Lambda^c}) \frac{e^{-\beta H_\Lambda(\sigma_\Lambda | \sigma_{\Lambda^c})}}{Z_\Lambda} \mu_\Lambda(d\sigma_\Lambda)$$

In other words, the regular conditional expectation is given by a skewed measure.

Note: we can use the tower rule on top of this!

$$\mathbb{E}[f(\omega)] = \mathbb{E}[\mathbb{E}[f(\sigma_\Lambda, \sigma_{\Lambda^c}) | \sigma_{\Lambda^c}]]$$

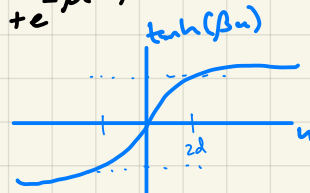
Example

Consider Ising model w/ ^{translation-invariant} $H = - \sum_{\langle x, y \rangle} \sigma_x \sigma_y - h \sum_x \sigma_x$
 We'd like to compute $\mathbb{E}[\sigma_0]$.

First, note that $H(\sigma_0 | \sigma_{\{0,3\}^c}) = - \sum_y \mathbb{1}_{0,y} \sigma_0 \sigma_y - h \sigma_0$
 $= - \sigma_0 \left[\sum_y \mathbb{1}_{0,y} \sigma_y + h \right]$

So, DLR with the uniform a priori measure μ gives

$$\mathbb{E}[\sigma_0 | \sigma_{\{0,3\}^c}] = \sum_{\sigma_0 \in \{\pm 1\}} \frac{\sigma_0 e^{\beta \sigma_0 (\sum_y \mathbb{1}_{0,y} \sigma_y + h)}}{\sum_{\sigma_0 \in \{\pm 1\}} e^{\beta \sigma_0 (\sum_y \mathbb{1}_{0,y} \sigma_y + h)}} = \frac{e^{\beta(\dots)} - e^{-\beta(\dots)}}{e^{\beta(\dots)} + e^{-\beta(\dots)}} = \tanh\left(\beta \left(\sum_y \mathbb{1}_{0,y} \sigma_y + h\right)\right)$$



By the tower rule, in any Gibbs stat,

$$\mathbb{E}[\sigma_x] = \mathbb{E}[\mathbb{E}[\sigma_x | \sigma_{\{0,3\}^c}]] = \mathbb{E}[\tanh(\beta (\sum_y \mathbb{1}_{0,y} \sigma_y + h))]$$

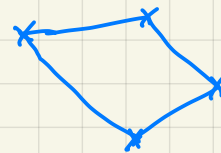
In nearest-neighbor interaction, $2d\beta < 1 \Rightarrow$ bounded tanh that allows us to converge the expanding region

We can (and eventually will) compute this outer expectation by repeatedly conditioning on a $\{0,3\}^c$. The, for β small enough, as we move further away, the bounds on tanh compound and things are nice.

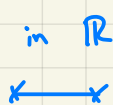
We would like to characterize the set of possible Gibbs measures for a certain (H, β) combination, as the coexistence of infinite Gibbs states is the hallmark of first order phase transitions! First, some vocabulary.

Def: The **extremal points** of a convex set K are the points $x \in K$ s.t. $\nexists a, b \in K, t \in (0, 1)$ s.t. $x = at + b(1-t)$

(basically the vertices)



Def: A **simplex** is a convex set K s.t. $\forall x \in K$, x has a unique representation as a convex sum (or integral) of the extremal points of K .



...

For finite $\dim K$, there are $|\dim K| + 1$ extremal points on the simplex, and all points $x \in K$ are expressible as a unique convex combination of them.

In infinite-dim K , all $x \in K$ are a unique integral over normalized measure, or an expectation.

Theorem: (Properties of infinite Gibbs measures)

For specified (H, β) , we have

- ① The set of Gibbs measures is closed under convex combination (and so the set is **convex**).
- ② In fact, the set of Gibbs measures is a **simplex**

Remarks:

- ① \Rightarrow if you have more than one Gibbs state, you have an infinite number
- ② \Rightarrow that every Gibbs state admits a unique extremal value decomposition.

Lecture 3/28.

Moving on, we now inspect the relationship between **uniqueness of Gibbs states** and **Symmetry breaking**.

Consider a probability space $(\Omega, \mathcal{F}, \mu)$. μ induces a Gibbs measure $\Delta(d\omega)$ and we have the DLR characterization of the Gibbs measure

$$\mathbb{E}_\Delta[f] = \mathbb{E}_\mu \left[\mathbb{E}_\Delta[f | \mathcal{F}_\Lambda^c] \right] = \int \mathbb{E}_\Delta[f | \mathcal{F}_\Lambda^c] \Delta(d\omega)$$

$$\text{with } \mathbb{E}_\Delta[f | \mathcal{F}_\Lambda^c] = \int_{\Omega_\Lambda} f(\omega_\Lambda, \omega_\Lambda^c) \frac{e^{-\beta H_\Lambda(\omega_\Lambda | \omega_\Lambda^c)}}{Z_\Lambda(\omega_\Lambda^c)} \mu(d\omega_\Lambda)$$

We have the following theorem:

Theorem: (limit exists as we condition at ∞)

Given a prob. space $(\Omega, \mathcal{F}, \mu)$ and a monotone decreasing sequence of sub- σ -algebras $\mathcal{F}_m \searrow$, then for any bounded measurable f

$$\mathbb{E}[f | \mathcal{F}_m](\omega) \xrightarrow[\text{a.s.}]{\text{to } \mu \text{ a.e.}} \mathbb{E}[f | \mathcal{F}_\infty](\omega)$$

$$\text{where } \mathcal{F}_\infty := \overline{\bigcap_n \mathcal{F}_n}$$

Proof: uses the martingale convergence theorem. Look it up \square

To apply this to our case,

$\mathcal{F}_\Lambda :=$ the σ -algebra of measurable sets of configurations within Λ

$\mathcal{B}_\Lambda :=$ the set of measurable functions induced by \mathcal{F}_Λ .

$$(f(\omega) \in \mathcal{B}_\Lambda \Rightarrow f \text{ depends only on } \Lambda)$$

We define $\mathcal{F}_\infty := \overline{\bigcap_{\Lambda \subset \mathbb{Z}^d, |\Lambda| < \infty} \mathcal{F}_\Lambda}$ and $\mathcal{B}_\infty := \bigcap_{\Lambda \subset \mathbb{Z}^d, |\Lambda| < \infty} \mathcal{B}_\Lambda$

Then, $f \in \mathcal{B}_\infty$ if f doesn't depend on the states inside any finite volume.

Examples

• An example $f \in \mathcal{B}_\infty$ is $f(\omega) := \limsup_{L \rightarrow \infty} \frac{1}{|\Lambda_L|} \sum_{x \in \Lambda_L} \omega_x$, $\Lambda_L := [-\frac{L}{2}, \frac{L}{2}]^d$

• Let $\mu_p(d\omega)$ be defined s.t. $\omega_i \sim \text{Bernoulli}(p)$ i.i.d.

The LLN $\Rightarrow \frac{1}{L} \sum_{i=1}^L \omega_i \xrightarrow{\text{a.s.}} p \Rightarrow f(\omega) \equiv p$ a.e. and \mathcal{F}_∞ is trivial ($\mathcal{F}_\infty = \{\emptyset, \Omega\}$)

For $\mu := \lambda \mu_{\frac{1}{2}} + (1-\lambda) \mu_{\frac{2}{3}}$, $f(\omega) = \begin{cases} \frac{1}{2} & \text{w.p. } \lambda \\ \frac{2}{3} & \text{w.p. } 1-\lambda \end{cases}$ and \mathcal{F}_∞ is not trivial!

Let $\mu(d\omega)$ be a Gibbs state. As before, but in the probabilistic notation,

$$\mathbb{E}_\mu[F] = \int \mathbb{E}[F | \Sigma_{\Lambda_c}](\omega) \mu(d\omega) \stackrel{\text{a.s.}}{=} \int \mathbb{E}[F | \Sigma_\infty](\omega) \mu(d\omega)$$

(Theorem) ↖ the measure really only matters at ∞

So, for μ -a.e. ω , $F(\omega) \mapsto \mathbb{E}[F | \Sigma_\infty](\omega)$ is a Gibbs measure since it satisfies DLR.

Theorem:

(i) Any Gibbs state can be presented as a convex combination of extremal Gibbs states.

(ii) A Gibbs state μ is **extremal** $\iff \Sigma_\infty$ is trivial w.r.t. μ (functions measurable at ∞ are constant a.e.; they're only supported on one type of configuration only)

Corollary:

If μ_1, μ_2 are extremal Gibbs states (for the same Hamiltonian), then either

(i) $\mu_1 = \mu_2$ or (ii) $\mu_1 \perp \mu_2$ (mutually singular; the measures are supported on different sets)

Lecture 3/30-

Infinite Gibbs States + Symmetry Breaking

Theorem (10.11):

Given an extensive Hamiltonian with finite energy per site and a Δ a sufficient condition for uniqueness of its infinite volume Gibbs state is:

for any pair of Gibbs measures Δ_1, Δ_2 , $\exists C < \infty$ s.t.
 \forall positive $f: \Omega \rightarrow \mathbb{R}_+$,

$$\mathbb{E}_{\Delta_1}[f] \leq C \mathbb{E}_{\Delta_2}[f] \quad (\text{absolutely continuous})$$

Proof: It suffices to prove that there exists a unique extremal Gibbs states.

This does not allow an $A \subseteq \Omega$ s.t. $\Delta_1(\mathbb{1}_A) = 1, \Delta_2(\mathbb{1}_A) = 0$

So, there cannot be two mutually-singular extremal Gibbs states, since otherwise there would be such an A . □

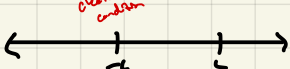
Ex/

$$H(\sigma) = -\sum_{x,y} J_{x,y} \sigma_x \sigma_y$$

Theorem:

For 1D arrays of (bounded) spins $\{\sigma_n\}$ with $\sum_{n \geq 0} |J_n| < \infty$, the Gibbs state is unique $\forall \beta < \infty$.

In particular, $J_n = \frac{1}{n^2}$ and $d \geq 2 \Rightarrow$ no 1st order phase transition.

Proof: Consider  For any x and y on opposite sides of i ,

$$\sum_{\substack{x \geq 0 \\ y < 0 \\ y < i}} |J_{x,y}| \leq \sum_{m \geq 1} |J_m| < C_0 < \infty. \quad \text{Now, for any } f \in \mathcal{B}_{[-1,1]},$$

$$\mathbb{E}_\Delta[f] \stackrel{\text{or}}{=} \int \mathbb{E}_\Delta[f | \sigma_{n_i}] \Delta(d\sigma_{n_i})$$

We have

$$\mathbb{E}_\Delta[f | \sigma_{n_i}] = \int f(\sigma) \frac{e^{-\beta H_L(\sigma_{n_i} | \sigma_{n_i^c})}}{Z_{n_i}} \Delta_{n_i}(d\sigma_{n_i})$$

and

$$H_L(\sigma_{n_i} | \sigma_{n_i^c}) = H_{n_i}(\sigma_{n_i}) + R_i(\sigma_{n_i}, \sigma_{n_i^c}) \quad \text{and} \quad |R_i(\sigma)| \leq 4C_0$$

$$\Rightarrow \mathbb{E}_\Delta[f | \sigma_{n_i}] \leq e^{4\beta C_0} \int f(\sigma_{n_i}) \frac{e^{-\beta H_{n_i}(\sigma_{n_i})}}{Z_{n_i}} \Delta_{n_i}(d\sigma_{n_i})$$

$$\geq e^{-4\beta C_0} \int f(\sigma_{n_i}) \frac{e^{-\beta H_{n_i}(\sigma_{n_i})}}{Z_{n_i}} \Delta_{n_i}(d\sigma_{n_i})$$

estimate doesn't depend on size of the volume!
we approximate measurable functions by local functions

same for any Δ_1, Δ_2 , call it $\mathbb{E}^{\text{ref}}[f | \sigma_{n_i}]$

$$\Rightarrow \mathbb{E}_{\Delta_1}[f|\sigma_{i,c}] \leq e^{4\beta C_0} \mathbb{E}^{\text{free}}[f|\sigma_{i,c}] \quad \text{and} \quad \mathbb{E}_{\Delta_2}[f|\sigma_{i,c}] \geq e^{-4\beta C_0} \mathbb{E}^{\text{free}}[f|\sigma_{i,c}]$$

$$\Rightarrow \mathbb{E}_{\Delta_1}[f] \leq e^{4\beta C_0} \int \underbrace{\mathbb{E}^{\text{free}}[f|\sigma_{i,c}]}_{\substack{\text{does not depend} \\ \text{on } \sigma_{i,c}}} \rho_1(d\sigma_{i,c}) = e^{4\beta C_0} \mathbb{E}_{\Delta_1}^{\text{free}}[f] \int \underbrace{\rho_1(d\sigma_{i,c})}_{=1}$$

$$\leq e^{8\beta C_0} \int \mathbb{E}_{\Delta_2}[f|\sigma_{i,c}] \rho_2(d\sigma_{i,c}) = e^{8\beta C_0} \mathbb{E}_{\Delta_2}[f]$$

By the previous theorem, we must have a unique Gibbs state. \square

How does this apply to continuous symmetry breaking?

Consider $H(\sigma) = - \sum_{x \sim y} \vec{\sigma}_x \cdot \vec{\sigma}_y$ with $\vec{\sigma}_x = (\sigma_x^1, \dots, \sigma_x^M)$, $\|\vec{\sigma}_x\| = 1$

and let $R_\theta \vec{\sigma} :=$ rotation of $\vec{\sigma}$ by θ .

Suppose a rotationally-invariant Hamiltonian and a-priori measure
 $H(\sigma) = H(R_\theta \sigma)$ and $\mu(d\sigma) = \mu(R_\theta d\sigma)$

Rotation on configuration space induces rotation on observable space by
 $(R_\theta f)(\sigma) := f(R_\theta \sigma)$

This, in turn, induces a rotation on the space of measures where
 $R_\theta \Delta$ is defined by $\mathbb{E}_{R_\theta \Delta}[f] = \mathbb{E}_{\Delta}[R_\theta f] \quad \forall f$

Then, we have two dichotomous options:

① either we have a unique, rotationally-invariant Gibbs state
 $\Delta = R_\theta \Delta \quad \forall \theta$

OR

② \exists an extremal Gibbs state Δ st. Δ and $R_\theta \Delta$ are singular
 - we can test for condition ② and discover when the

R_θ -symmetry of Δ is broken

Renormalization group hub?
 the boundary $\Gamma \sim \frac{1}{|x-y|^2}$ model


$$\{x, y\} \quad H(x, y) = \int_{x, y} \sigma_x \sigma_y \quad \sigma_x = (\sigma_x^1, \sigma_x^2) \quad \|\sigma_x\| = 1$$

The boundary case: $\sigma_x = \pm 1$

$$J_{xy} = \frac{1}{|x-y|^2} \quad x \neq y$$

$$\sum_{x \neq 0} |J_{0x}| < \infty$$

$$\sum_{y \neq 0} |J_{0y}| = \infty$$



$\langle \sigma_x^+ \rangle = M$
 $\sum_D = \frac{1}{|D|} \sum_{x \in D} \sigma_x$
 $\sum_{B_m} = \pm 1$
 $\sum_{x \in B_m} \int_{y \in B_m} \sigma_x \sigma_y = \sum_{x \in B_m} \sum_{y \in B_m} \sigma_x \sigma_y$
 $L^2 \frac{1}{(mL)^2} = \frac{1}{m^2}$

Lecture 4/4 - Mermin-Wagner Theorem

First we ought to verify that a rotated infinite-volume Gibbs state is still an infinite-volume Gibbs state: the DLR condition verifies this.

Theorem: (Mermin-Wagner)

For a two dimensional finite-range system of continuous spin variables with rotational symmetry, i.e.

$$H(\sigma) = - \sum_{A \subset \Lambda} \int_A \phi_A(\sigma_A),$$

$\text{diam}(A) \leq R$

if (i) ϕ_A is invariant under uniform rotations and (ii) ϕ_A varies smoothly under all rotations (ex: $H = - \sum_{x,y} J_{xy} \vec{\sigma}_x \cdot \vec{\sigma}_y$)

then any infinite-volume Gibbs state is invariant under uniform spin rotations; i.e. \forall observables $f: \mathbb{R} \rightarrow \mathbb{R}$ and all $\mathcal{R}_\theta \in O(2)$,

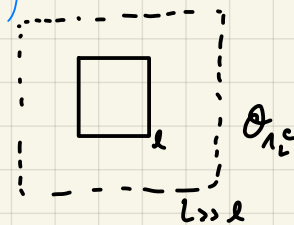
$$\int f d\mu = \int \mathcal{R}_\theta f d\mu \iff \mathbb{E}_\Delta[f] = \mathbb{E}_\Delta[\mathcal{R}_\theta f]$$

Proof: It suffices to show that \forall extremal Gibbs states Δ , $\exists c < \infty$ s.t. \forall local $f: \mathbb{R} \rightarrow \mathbb{R}$ with $f \geq 0$,

$$\mathbb{E}_\Delta[\mathcal{R}_\theta f] \leq c \mathbb{E}_\Delta[f] \quad (\Delta \text{ absolutely continuous to } \mathcal{R}_\theta \Delta \ \forall \Delta)$$

Fix a volume Λ_L s.t. $f \in \mathcal{B}_{\Lambda_L}$. For any larger Λ_L , the tower rule and DLR condition give

$$\mathbb{E}_\Delta[f] = \mathbb{E}_\Delta[\mathbb{E}_\Delta[f | \mathcal{O}_{\Lambda_L^c}]]$$



Consider a soft, nonuniform rotation of spins given by angle

$$\theta(x) = \begin{cases} \theta & |x| \leq L \\ 0 & |x| > L \end{cases}$$

and a rotation operator on functions $\tilde{\mathcal{R}} f(\sigma) := f(\{\mathcal{R}_{\theta(x)} \sigma_x\}_{x \in \Lambda})$

Then,

$$\begin{aligned} \mathbb{E}_\Delta[\tilde{\mathcal{R}} f | \mathcal{O}_{\Lambda_L^c}] &\stackrel{\text{DLR}}{=} \int_{\mathcal{O}_{\Lambda_L}} \tilde{\mathcal{R}} f(\sigma) \frac{e^{-\beta H_\Delta(\sigma | \mathcal{O}_{\Lambda_L^c})}}{Z_\Delta(\sigma | \mathcal{O}_{\Lambda_L^c})} \mu(d\sigma) \\ &= \int f(\tilde{\mathcal{R}}\sigma) \frac{e^{-\beta [H_\Delta(\sigma | \mathcal{O}_{\Lambda_L^c}) - H_\Delta(\tilde{\mathcal{R}}\sigma | \mathcal{O}_{\Lambda_L^c})]}}{Z_\Delta(\sigma | \mathcal{O}_{\Lambda_L^c})} \frac{e^{-\beta H_\Delta(\tilde{\mathcal{R}}\sigma | \mathcal{O}_{\Lambda_L^c})}}{Z_\Delta(\tilde{\mathcal{R}}\sigma | \mathcal{O}_{\Lambda_L^c})} \mu(d\tilde{\mathcal{R}}\sigma) \end{aligned}$$

weighting factor based on effect of $\tilde{\mathcal{R}}$ on the energy

$\tilde{\mathcal{R}}$ is bijective and product measure, $\mu(d\tilde{\mathcal{R}}\sigma) = \mu(d\sigma)$

We have

$$\begin{aligned} H_\Delta(\tilde{\mathcal{R}}\sigma) - H_\Delta(\sigma) &\stackrel{\text{energy penalty}}{=} - \sum_{|x-y|=1} \int_{x,y} [\mathcal{R}_{\theta(x)} \vec{\sigma}_x \cdot \mathcal{R}_{\theta(y)} \vec{\sigma}_y - \vec{\sigma}_x \cdot \vec{\sigma}_y] \\ &= - \sum_{|x-y|=1} \int_{x,y} [\vec{\sigma}_x \cdot (\mathcal{R}_{\theta(x)} \vec{\sigma}_y - \vec{\sigma}_y)] \end{aligned}$$

We would like to bound $\vec{\sigma}_x \cdot (\mathcal{R}_{\Delta 0} \vec{\sigma}_y - \vec{\sigma}_y)$ to show that the energy penalty is favorable.

For $\vec{\sigma}_x \approx \vec{\sigma}_y$, there is only a second-order term

$$\vec{\sigma}_x \cdot (\mathcal{R}_{\Delta 0} \vec{\sigma}_y - \vec{\sigma}_y) \leq \frac{1}{2} |\Delta|^2$$

Otherwise, we require a linear term that we must bound with trickery.

Corollary:

In the above setting, there can be no spontaneous magnetization.
In other words, $\mathbb{E}_{\Delta}[\vec{\sigma}_x] = 0 \quad \forall x$.

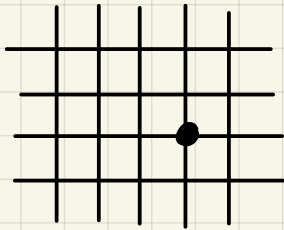
Proof: By Mermin-Wagner, $\mathbb{E}_{\Delta}[\vec{\sigma}_x] = \mathbb{E}_{\Delta}[-\vec{\sigma}_x]$. The result follows. \square

Cycles in random permutations Schramm

Lecture 4/11 - Q-State Potts

Q-State Potts Model

Consider a setup on a graph $G=(V,E)$ with



$$\Omega_0 = \{1, \dots, Q\}$$

$$Z(\beta, J) = \frac{e^{-\beta H(\sigma)}}{Z}$$

$$H = - \sum_{\langle x,y \rangle \in E} J_{x,y} (\underbrace{\delta_{\sigma_x \sigma_y}}_{\text{check this!}} - 1)$$

We call this the **Q-state Potts model**.

(When $Q=2$, we have Ising-like $\Omega_0 = \{1, 2\} \Rightarrow \delta_{\sigma_x \sigma_y} = \frac{\sigma_x \sigma_y + 1}{2}$ in the Ising case, so, $Q=2$ is scaled + shifted Ising)

So,

$$e^{\beta J_{x,y} (\delta_{\sigma_x \sigma_y} - 1)} = \sum_{\sigma_x, \sigma_y} \left(1 - e^{-\beta J_{x,y}} \right) + e^{-\beta J_{x,y}} = p_{x,y} \delta_{\sigma_x \sigma_y} + (1 - p_{x,y}) \cdot 1$$

$$\Rightarrow Z = \sum_{\sigma} \prod_{e \in E} \left[p_{x,y} \delta_{\sigma_x \sigma_y} + (1 - p_{x,y}) \cdot 1 \right] = \sum_{\sigma: V \rightarrow \{1, \dots, Q\}} \prod_{e \in E} p_e^{n(e)} (1 - p_e)^{1 - n(e)} \prod_{u \in V} \delta_{\sigma_u}$$

sum over random clusters where $n(e) = \mathbb{1}_{\{e \text{ is active}\}}$
 \sim Bernoulli: (p_e) if spins agree, 0 otherwise
 $= \mathbb{1}_{\{\sigma \text{ is constant on cluster } n\}}$

$$\Rightarrow Z = \sum_{n: E \rightarrow \{0,1\}} \prod_{\langle x,y \rangle \in E} p_{x,y}^{n(\langle x,y \rangle)} (1 - p_{x,y})^{1 - n(\langle x,y \rangle)} Q^{N_c(n)}$$

of connected clusters where $p_{x,y} = 1 - e^{-\beta J_{x,y}}$

Spin-spin correlation

Given a cluster n , we have

$$E[\delta_{\sigma_u \sigma_v} | n] = \begin{cases} 1 & \text{if } n(\{u,v\}) = 1 \\ \frac{1}{Q} & \text{else} \end{cases}$$

depends on u,v only on if n is same cluster!

From here, we can see that the Q-state Potts model forms clusters with the same σ_u 's that flip together. This makes it a perfect place to study percolation.

If $Q=2$, we have $\langle \sigma_u \sigma_v \rangle = n(\{u,v\})$ with σ as in Ising model.

Remarks:

- $Q=1$ is random percolation model
- $Q=2$ is scaled/shifted Ising
- If we take $Q \in \mathbb{R}^+$ instead, we get the **F-K random cluster model**
- As $Q \rightarrow 0$, measure concentrates around clusters with low N_c ; i.e. \rightarrow minimum spanning tree

When we consider the connection of an edge, conditioned on the rest of n (i.e. $\mathbb{P}\{n(e)=1 \mid \{n(e') : e' \neq e\}\}$),

$$\frac{\mathbb{P}\{n(e)=1 \mid \begin{array}{c} v \\ \text{---} \\ u \end{array}\}}{\mathbb{P}\{n(e)=0 \mid \begin{array}{c} v \\ \text{---} \\ u \end{array}\}} = \frac{p}{1-p}$$

$$\frac{\mathbb{P}\{n(e)=1 \mid \begin{array}{c} v \\ \text{---} \\ u \end{array} \times\}}{\mathbb{P}\{n(e)=0 \mid \begin{array}{c} v \\ \text{---} \\ u \end{array} \times\}} = \frac{\frac{p}{p+(1-p)Q}}{\frac{(1-p)Q}{p+(1-p)Q}} = \frac{p}{(1-p)Q} \leq p$$

In the dual, 0 and 1 flip and the arrows swap. So, the model is **self-dual** when

$$\frac{p}{1-p} = \frac{(1-p)Q}{p} \Leftrightarrow \frac{p}{1-p} = \sqrt{Q}$$

This is the **Kramers-Wannier self-duality point** in 2D.

FKG Monotonicity (Fortuin-Kostlyn-Ginibre)

The collection of possible clusters $\eta: E \rightarrow \{0,1\}$ (which we denote $\{0,1\}^E$) is partially ordered.

Def: (partial ordering)

An ordering $>$ is a **partial ordering** if

(i) $n' > n \iff n'_e \geq n_e \quad \forall e$

(ii) $f: \{0,1\}^E \rightarrow \mathbb{R}$ is \uparrow if $f(n') \geq f(n) \quad \forall n' > n$

(iii) For prob. measures μ_1, μ_2 on $\{0,1\}^E$,

$$\mu_1 > \mu_2 \iff \forall f \uparrow, \int f(n) \mu_1(dn) \geq \int f(n) \mu_2(dn)$$

" μ_1 dominates μ_2 "

not like \mathbb{Z}^d but in same direction

Def: A partially-ordered set forms a **"lattice"** iff \forall pairs (n', n) , there exists $n \vee n', n \wedge n'$ s.t. $n \vee n' > n, n'$ and $n \wedge n' < n, n'$

In this case, $(n \vee n')(e) = \max\{n(e), n'(e)\}$, $(n \wedge n')(e) = \min\{n(e), n'(e)\}$

Def: A probability measure μ is **positively associated** if $\forall f, g: \Omega \rightarrow \mathbb{R}$ s.t. $f, g \geq 0$ and $f, g \uparrow$, we have

$$\mathbb{E}_\mu[f g] \geq \mathbb{E}_\mu[f] \mathbb{E}_\mu[g]$$

"monotonic, nonnegative functions are positively associated"

check this definition

Theorem

For two measures μ_1, μ_2 on $\{0,1\}^E$,

$\mu_1 > \mu_2 \iff$ there exists a coupling $\mu(dn_1, dn_2)$ s.t.

(i) $\int g(n_j) \mu(dn_1, dn_2) = \int g(n) \mu_j(dn) \quad j=1,2$ (correct marginals)

(ii) $n_1 > n_2$ μ -a.s.

Note that for $f \uparrow$, the second condition implies

$$\mathbb{E}_{\mu_1}[f] - \mathbb{E}_{\mu_2}[f] = \int \underbrace{[f(n_1) - f(n_2)]}_{\geq 0} \mu(dn_1, dn_2) \geq 0$$

Theorem:

Let μ be a probability measure on a partially-ordered "lattice".
 A sufficient condition for μ to have positive association is that

$$\mathbb{E}_\mu[n \vee n'] \mathbb{E}_\mu[n \wedge n'] \geq \mathbb{E}_\mu[n] \mathbb{E}_\mu[n'] \quad \forall n, n'$$

Example (Ising)

$$\sigma' \succ \sigma \Leftrightarrow \sigma'_x \geq \sigma_x \quad \forall x \quad \text{and} \quad \Delta(\sigma) = \frac{e^{-\beta \sum_{x,y} J_{xy} \sigma_x \sigma_y}}{Z}$$

Consider

σ	+ - + - - -
σ'	- - - + + -
$\sigma \vee \sigma'$	+ - + + + -
$\sigma \wedge \sigma'$	- - - - - -

these states are more likely than the originals, since now spin agreement

(We can write the relation $(\sigma'_x \wedge \sigma_x)(\sigma'_y \wedge \sigma_y) + (\sigma'_x \vee \sigma_x)(\sigma'_y \vee \sigma_y) \geq \sigma'_x \sigma'_y + \sigma_x \sigma_y$) ← not sure what this has to do w/ anything

So, Ising spin model Gibbs measure Δ satisfies the theorem, since $\sigma \wedge \sigma'$ and $\sigma \vee \sigma'$ are more likely than σ or σ' .

Example (FK random cluster model)

The relation $\mathbb{E}_\mu[n \vee n'] \mathbb{E}_\mu[n \wedge n'] \geq \mathbb{E}_\mu[n] \mathbb{E}_\mu[n']$ holds iff

$$\frac{\mu(n' \cup \{e\})}{\mu(n')} \geq \frac{\mu(n \cup \{e\})}{\mu(n)} \quad \forall n, n' \text{ s.t. } n'(e) \geq n(e) \quad \forall e \neq \emptyset$$

We can verify this for FK random cluster model.

Example (Q-state Potts model)

Note that the Gibbs measure $\Delta_{\beta, Q}(n) = \prod_{e \in E} p_{xy}^{n(e)} (1-p_{xy})^{1-n(e)} Q^{N_c(n)}$

has that

- $\Delta_{\beta, Q}$ is decreasing in Q
- $N_c(n)$ is decreasing in n
- $\Delta_{\beta, Q}$ is increasing in β
- $N_c(n) + |n|$ is increasing in n

Also, $\forall Q' \geq Q \geq 1$

$$\beta_c(Q') \geq \beta_c(Q) \geq \frac{Q}{Q'} \beta_c(Q')$$

This relates critical points of models for different Q 's! So, critical behavior in one implies critical behavior in another!

Interpretation:

For any $\Delta(d_n)$ satisfying the FKG condition, $\forall f, g \geq 0$ with $f, g \nearrow$
positive associativity gives

$$\mathbb{E}_\Delta[gf] \geq \mathbb{E}_\Delta[g] \mathbb{E}_\Delta[f] \Rightarrow \int g(n) f(n) \mu(d_n) \geq \left(\int g(n) \mu(d_n) \right) \left(\int f(n) \mu(d_n) \right)$$

$$\Rightarrow \frac{\int g(n) f(n) \Delta(d_n)}{\int f(n) \Delta(d_n)} \geq \int g(n) \Delta(d_n)$$

So, letting $\mu(d_n)$ be the tilted measure $\mu(d_n) := f(n) \Delta(d_n)$, then
 $\mu \succ \Delta$

Holley's Theorem:

We have that $\Delta < \Delta'$ if and only if there exists a coupling $\mu(d_x, d_{x'})$ s.t. μ has marginal distributions agreeing with Δ and Δ' and μ is supported only on states $\sigma < \sigma'$.

The above theorem grants that if $\Delta < \Delta'$, then

$$\Delta'(\sigma_x) - \Delta(\sigma_x) = 2\mu(\sigma_x \neq \sigma_{x'}) \leftarrow \text{sometimes wrong w/ } \mu \text{!}$$

Add

Lecture

4/18

Lecture 4/20-

Quantum Spin Chains

Suppose we have a Hamiltonian \hat{H} that we have diagonalized.

We have the \mathcal{Q} -thermal expectation value

$$\hat{A} \mapsto \langle \hat{A} \rangle_{\beta} = \frac{\text{tr}(\hat{A} e^{-\beta \hat{H}})}{\text{tr}(e^{-\beta \hat{H}})}, \quad \text{with} \quad Z := \text{tr}(e^{-\beta \hat{H}}) = \sum_{n=1}^{\infty} \langle \psi_n | e^{-\beta \hat{H}} | \psi_n \rangle$$

$\{\psi_n\}_n$ is any ONB

We can label Hamiltonians artificially by the time t_j at which we added it to the picture, giving a Dyson integral

$$e^{-\beta \hat{H}} = \sum_{n=0}^{\infty} \int \dots \int_{0 \leq t_1, \dots, t_n < \beta} (-\hat{H})_{t_n} \dots (-\hat{H})_{t_1} dt_1 \dots dt_n, \quad \text{with} \quad \hat{H}_t \equiv \hat{H}$$

Writing $-\hat{H} = \sum_{\alpha \in \mathcal{I}} \hat{K}_{\alpha}$, we get

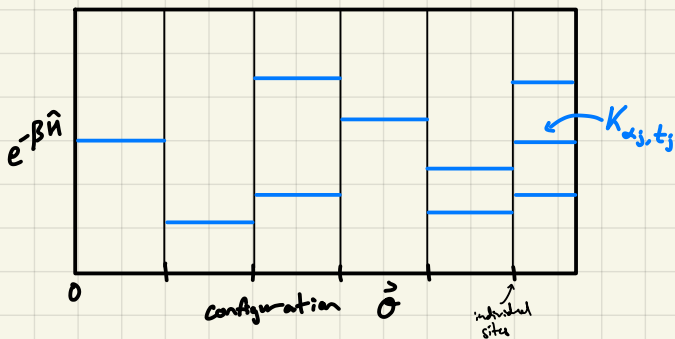
$$e^{-\beta \hat{H}} = \sum_{n=0}^{\infty} \int \dots \int_{0 \leq t_1, \dots, t_n < \beta} \sum_{\{\alpha_1, \dots, \alpha_n\} \subseteq \mathcal{I}} \left(\prod_{j=1}^n \hat{K}_{\alpha_j t_j} \right) dt_1 \dots dt_n$$

$$= e^{\beta |\mathcal{I}|} \int_{\Omega} \prod_{j=1}^{|\mathcal{I}|} \hat{K}_{\alpha_j t_j} \Delta(d\omega)$$

\leftarrow time-ordered product

(note that the Dyson integral came from $e^{i\hat{H}t}$. we apply analytic continuation from imaginary time to be in real time)

In a sense, we have written out a random formulation for \hat{H} with the following picture



In this picture, we visualize quantum evolution (in real time) as randomly applying operators $\hat{K}_{\alpha_j t_j}$ as we move through (inverse) time. We assign measure $\Delta(d\omega)$ to the sequence of these transformations.

$$\mathcal{I} \times [0, \beta]$$

$$\omega = \{\alpha_j, t_j\}_{j=1}^{|\mathcal{I}|}$$

$\Delta(d\omega)$ is a Poisson process on $\mathcal{I} \times [0, \beta]$

More on \mathcal{I} vs \mathcal{E}

Note that our integrals will look like the usual d -dimensional stuff + 1 extra dimension for the imaginary time! So, quantum stat mech in d -dim will look like classical stat mech in $d+1$ -dim.

$$Z_{\mathcal{I}, \beta} = e^{\beta |\mathcal{I}|} \int_{\Omega} \text{tr} \left(\prod_{j=1}^{|\mathcal{I}|} \hat{K}_{\alpha_j t_j} \dots \hat{K}_{\alpha_1 t_1} \right) \Delta(d\omega)$$

Q - Spin Operators

Hermitian
(self-adjoint)

We have $\hat{S} = (\hat{S}_x, \hat{S}_y, \hat{S}_z)$ satisfying $[\hat{S}_x, \hat{S}_y] = i\hat{S}_z$ with cyclic permutation.

Note that $|\hat{S}|^2 = \hat{S}_x^2 + \hat{S}_y^2 + \hat{S}_z^2$ commutes with each of $\hat{S}_x, \hat{S}_y, \hat{S}_z$.
This is because the magnitude of \hat{S} is invariant under rotations $e^{i\theta \hat{S}_z}$.

We can also derive $|\hat{S}|^2 = s(s+1)\mathbb{1}$

Suppose our states live in a finite-dim Hilbert space \mathcal{H} .

If $\dim \mathcal{H} = N$, we get an ONB for \mathcal{H} from eigenvectors of \hat{S}_z

$$\hat{S}_z |s, m\rangle = m |s, m\rangle, \quad m \in \{-s, -s+1, \dots, s-1, s\}, \quad s = \frac{N-1}{2}$$

s can be half-integer

(Fun story: we can prove the quantization for \hat{S}_z by slicing the surface of a sphere into slices. In 3D, Archimedes proved equally-spaced slices have equal area, implying integer quantization)

Suppose we have two spins, modeled as

$$\begin{aligned} \mathcal{H}_{s_1} \otimes \mathcal{H}_{s_2} &= \text{span}\{|s_1, m_1\rangle \otimes |s_2, m_2\rangle\}_{m_1, m_2} \\ &= \bigoplus_{|s_1-s_2|}^{s_1+s_2} \mathcal{H}_{s_{1,2}} \end{aligned}$$

$\mathcal{H}_{s_{1,2}}$ group representation of $\mathcal{H}_{s_1} \otimes \mathcal{H}_{s_2}$

Let $s_{1,2}$ be the possible magnitudes of the combined spin, i.e. the values

$$s_{1,2} \in \{|s_1-s_2|, \dots, s_1+s_2\} \Rightarrow \hat{S} = \hat{S}_1 + \hat{S}_2 = \begin{matrix} |s_1-s_2| \\ \vdots \\ s_1+s_2 \end{matrix} \left(\begin{array}{c} \square \\ \square \\ \vdots \\ \square \end{array} \right)$$

Each of these subspaces are the $\mathcal{H}_{s_{1,2}}$'s

Examples

① Q-bit $s = \frac{1}{2}$, $\dim \mathcal{H} = 2$

We can write $\hat{S} = \frac{1}{2} (\sigma_x, \sigma_y, \sigma_z)$ with Pauli matrices

$$\sigma_x = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \quad \sigma_y = \begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix}, \quad \sigma_z = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$

Writing the Heisenberg anti-ferromagnetic/ferromagnetic spin chain on \mathbb{Z}

$$H = \pm 2 \sum_n \hat{\sigma}_n \cdot \hat{\sigma}_{n+1} \Rightarrow \hat{H} = \pm \sum_n \hat{S}_n \cdot \hat{S}_{n+1}$$

+ for anti-ferromagnetic

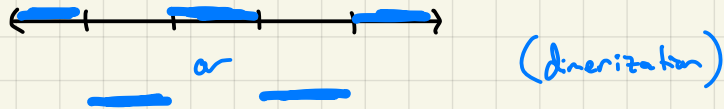
If they are all identical copies, we can write

$$\hat{S}_1 \cdot \hat{S}_2 = \frac{1}{2} [|\hat{S}_1 + \hat{S}_2|^2 - \hat{S}_1 \cdot \hat{S}_1 - \hat{S}_2 \cdot \hat{S}_2] = \frac{1}{2} [S_n(S_n+1) - 2S(S+1) \mathbb{1}]$$

For $s = \frac{1}{2}$ spins, $s_{1,2} \in \{0, 1\}$. Let $|\psi\rangle = \frac{|+-\rangle - |-+\rangle}{\sqrt{2}}$, and define $\hat{P}_{uv}^{(s)} := |\psi\rangle\langle\psi|$

Then, $(\hat{S}_{z,1} \hat{S}_{z,2}) |\psi\rangle = -|\psi\rangle$, and $\hat{S}_1 \cdot \hat{S}_2 = \frac{1}{2} - 2 \hat{P}_{uv}^{(s)}$ do the algebra (projection onto singlet)

We can't have all links in their lowest states, and so we can assign an initial ground state



Depending on evenness or oddness of the finite volume, one ground state will be preferable to another; translational symmetry breaking can occur if the presence of these two ground states remains in the infinite limit.

Lecture 4/25 - Quantum Spin Models

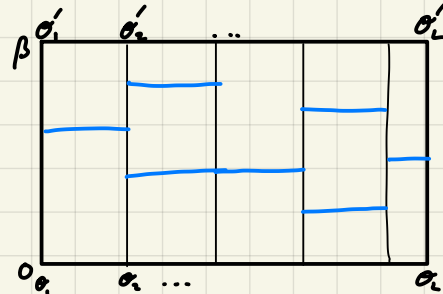
Recall the setup with

$$\hat{H} = - \sum_n \hat{S}_n \cdot \hat{S}_{n+1} =: \sum_{\alpha \in I} \hat{K}_\alpha$$

where $\hat{S}_j = \frac{1}{2}(\sigma_j^x, \sigma_j^y, \sigma_j^z)$ are Pauli matrices and $|\sigma_j\rangle$ denote eigenvectors of σ_j^z .

To compute expectations of the form $\langle \hat{A} \rangle_{\Lambda, \beta} = \frac{\text{Tr}(e^{-\beta \hat{H}} \hat{A})}{\text{Tr}(e^{-\beta \hat{H}})}$, we use the expansion

$$e^{\beta \sum_{\alpha \in I} \hat{K}_\alpha} = \sum_{n \geq 0} \int \dots \int_{0 \leq t_1, \dots, t_n \leq \beta} \left(\prod_{j=1}^n \hat{K}_{\alpha_j t_j} \right) dt_1 \dots dt_n$$



$w = \{(t_j, \tau_j)\}_j \subseteq I \times [0, \beta]$
poisson process in dtd-dim spacetime

To convert this to a probability measure, we normalize by dividing by $e^{\beta |I|}$. So, we get

$$e^{\beta \sum_{\alpha \in I} \hat{K}_\alpha} = e^{\beta |I|} \int_{\Omega} \left(\prod_{(x,t) \in w} \hat{K}_{x,t} \right) \Delta_{\beta, \Lambda}(dw)$$

If we want matrix elements in $\mathcal{H} := \text{span}\{|\sigma_1, \dots, \sigma_L\rangle\}$

$$\langle \sigma' | e^{\beta \sum_{\alpha \in I} \hat{K}_\alpha} | \sigma \rangle = e^{\beta |I|} \int_{\Omega} \langle \sigma' | \prod_{(x,t) \in w} \hat{K}_{x,t} | \sigma \rangle \Delta_{\beta, \Lambda}(dw)$$

$$\Rightarrow Z_{\Lambda, \beta} = \text{Tr}(e^{-\beta \hat{H}}) = \int_{\Omega} \sum_{\sigma} \langle \sigma' | \prod_{(x,t) \in w} \hat{K}_{x,t} | \sigma \rangle \Delta_{\beta, \Lambda}(dw)$$

over on $\Omega \times \mathcal{H}$

Returning to the **Heisenberg Anti-Ferromagnet**,

$$\hat{H} = + \sum_n \hat{S}_n \cdot \hat{S}_{n+1} = \sum_n -4 \hat{P}_{n, n+1}^{(0)} + \text{constant}$$

projection onto the singlet $|\psi\rangle = \frac{1}{\sqrt{2}}(|\uparrow\rangle|\downarrow\rangle - |\downarrow\rangle|\uparrow\rangle)$, $\text{proj}_{|\psi\rangle}(\psi)$

If we wished to extend to other spin values than spin-1/2, we can either leave the Hamiltonian as $\hat{S}_n \cdot \hat{S}_{n+1}$. Alternatively, we can write the singlet as (with s=1/2, people define spin systems in order to get certain properties)

$$(s \in \frac{1}{2} \mathbb{Z}_+) \quad |\psi\rangle := \frac{1}{\sqrt{2s+1}} \sum_{m=-s}^s (-1)^m |m, -m\rangle \Rightarrow \hat{P}^{(0)} = |\psi\rangle\langle\psi|$$

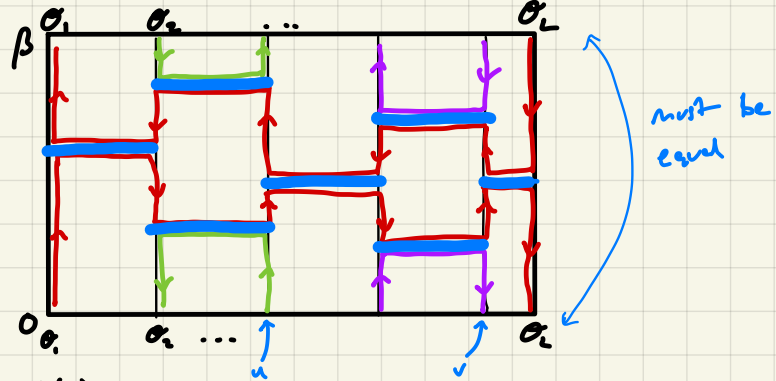
So, we consider $\hat{H} = - \sum_n \underbrace{(2s+1)}_{k_n} \hat{P}_{n, n+1}^{(0)} \quad (s \in \frac{1}{2} \mathbb{Z}_+)$

This gives matrix elements

$$\langle \sigma'_n \sigma'_{n+1} | \hat{K}_n | \sigma_n \sigma_{n+1} \rangle = \sum_{m, m'=-s}^s (-1)^{m-m'} \mathbb{1}[\sigma_n = -\sigma_{n+1} = m] \mathbb{1}[\sigma'_n = -\sigma'_{n+1} = m']$$

We know that spins either align or are opposite. However, the form of this problem forces neighboring spins to be opposite.

Note that if $\sigma = \sigma$ (as it does when computing traces), we can emphasize this constraint in the picture via:



Let N_L be the number of loops (3 in the depicted picture). Then, we have a degree of freedom of choice for each loop, and for each choice we can select $(2s+1)$ options. So, in this case,

$$Z_{1,\beta} = e^{\beta L} \int_{\Omega} (2s+1)^{N_L(\omega)} \Delta(d\omega)$$

Let's compute a spin-spin correlation.

$$\langle \hat{S}_u \cdot \hat{S}_v \rangle_{L,\beta} = 3 \langle \hat{S}_u^z \hat{S}_v^z \rangle_{L,\beta} = \frac{3 \text{Tr}(e^{-\beta \hat{H}} \hat{S}_u^z \hat{S}_v^z)}{\int_{\Omega} (2s+1)^{N_L(\omega)} \Delta(d\omega)} = \frac{e^{\beta L} \int_{\Omega} \sum_{\sigma} \langle \sigma | \prod_{(u,v) \in \omega} \hat{K}_{(u,v)} \hat{S}_u^z \hat{S}_v^z | \sigma \rangle \Delta_{\beta,\Lambda}(d\omega)}{\int_{\Omega} (2s+1)^{N_L(\omega)} \Delta(d\omega)}$$

If u and v are not connected by ω , they will average to 0 as \hat{S}_u^z, \hat{S}_v^z commute with the \hat{K} 's. If they are connected, their value will be determined by the loop, and will be

$$\left(\frac{1}{2s+1} \sum_{m=-s}^s m^2 \right) \cdot (-1)^{u-v} \leftarrow \text{spin alternate}$$

$:= M_s^2$

$$\implies \langle \hat{S}_u \cdot \hat{S}_v \rangle_{L,\beta} = 3 \langle \hat{S}_u^z \hat{S}_v^z \rangle_{L,\beta} = 3 (-1)^{u-v} M_s^2 \mathbb{P}\{u, v \text{ connected by } \omega\}$$

\uparrow in same loop

Note the relation with the Q-state Potts Model, whose partition function and correlations are analogous as follows:

	quantum spin chain	Q-state Potts
dim	d	$d+1$
part. fn.	number of loops	number of clusters
correlations	u, v in same loop	u, v in same cluster

Infinite-width

If we take $L \rightarrow \infty$ (with fixed parity of L), we can use FK G to show that the infinite measure converges and is invariant under translations by even shifts.

For different choices of d and s , we can get different results for uniqueness of Gibbs states, correlation decay, etc.

A Dichotomy for 2D Loop Systems



In the infinite limit, either every point is contained in infinitely many loops or all points are in finitely many loops. In the finite case, the parity of the loops introduces **dimerization**, causing long range order and **trichotomy symmetry breaking**.

Consider the following derivation: by translation invariance,

$$\sum_n n |\langle \hat{S}_0 \cdot \hat{S}_n \rangle| = \sum_{\substack{u>0 \\ v<0}} |\langle \hat{S}_u \cdot \hat{S}_v \rangle| = M_s^2 \sum_{\substack{u>0 \\ v<0}} \mathbb{P} \left\{ \left\langle \overset{\text{canceled by}}{\leftarrow} \underset{0}{\leftarrow} \overset{\text{or}}{\rightarrow} \right\rangle \right\}$$

Since loops don't overlap in Hershberg's antiferromagnet, every loop containing the origin must add another connected $u-v$ pair. So,

$$\sum_n n |\langle \hat{S}_0 \cdot \hat{S}_n \rangle| \geq M_s^2 \mathbb{E}[\# \text{ loops encircling } 0]$$

If # of loops about 0 is finite (Kolmogorov 0-1 gives $\mathbb{E}[\#] = \infty$), then the sum must also diverge. In particular, $\langle \hat{S}_0 \cdot \hat{S}_n \rangle$ decays no quicker than $\frac{1}{n^2}$.

So, we get that either

(i) dimerization + trichotomy symmetry breaking + long range order (multiple Gibbs measures, shifted by 1)

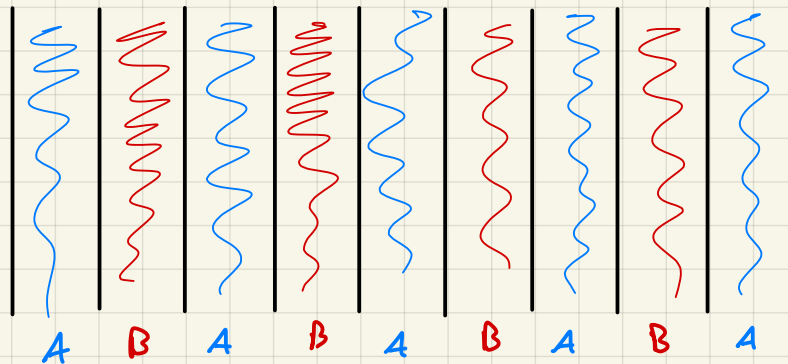
(ii) spin-spin correlation decays slower than $\frac{1}{|n|^{1+\epsilon}}$

This is a result of a general result: **2D loop dichotomy**

either (i) long-range-order or (ii) slower correlation decay

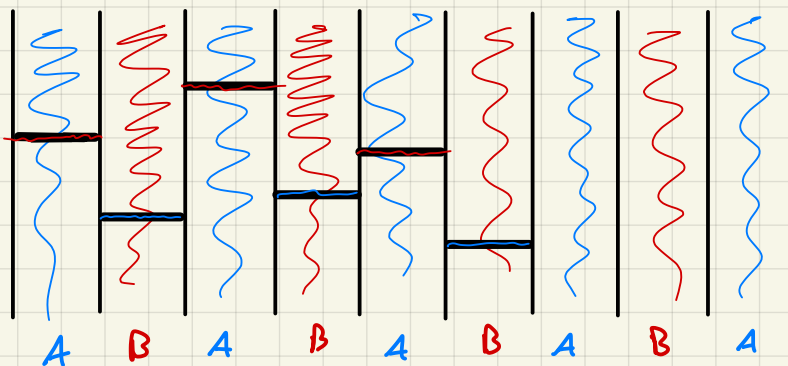
Lecture 4/27 - Final Lecture

We will now discuss the **A/B continuum percolation model**.
 If the space is split into strips,



"consider the city of"
 Venice

Now consider randomly placed connections between strips

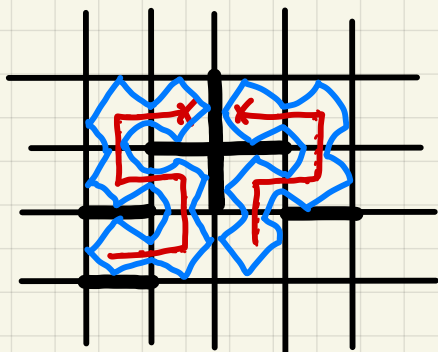


Each rung "connects"
 strips of the same
 color.

If we consider A as (+) and B as (-), we can place a total order on states via $w < w' \iff$ under w' , A is "more connected"

This arises naturally from investigation into quantum spin models.

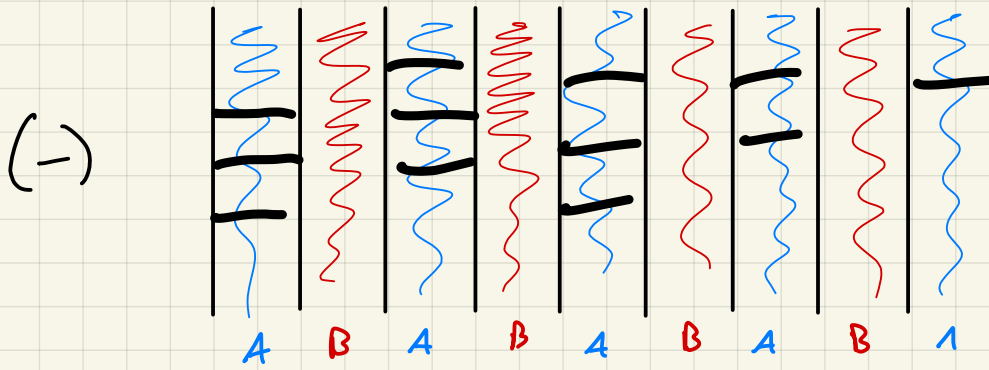
However, recall **FK-random cluster models** (Q -state Potts w/ $Q \in \mathbb{R}^+$)



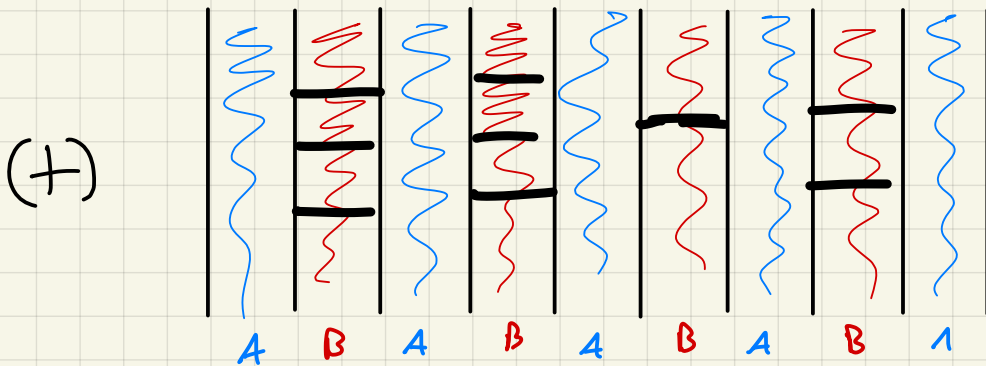
We can show that as we take the continuum limit of the lattice, we get a model equivalent to the A/B above, with the difference that the A/B measure is weighted by an extra factor

$$(2s+1)^{N_c(w)} \leftarrow \# \text{ of clusters}$$

Note that each time we add a ring to the same color, we add another cluster. In the extreme, we find two preferred types of states



and



In the infinite limit, we can find that both

$$\psi^- = \lim_{\substack{L \rightarrow \infty \\ L \text{ even}}} \frac{1}{Ld} \log(\text{Tr} e^{-\beta H}) \quad \text{and} \quad \psi^+ = \lim_{\substack{L \rightarrow \infty \\ L \text{ odd}}} \frac{1}{Ld} \log(\text{Tr} e^{-\beta H})$$

When $s > \frac{1}{2}$, $\psi^- \neq \psi^+$, and there is symmetry breaking via dimensionality.
 When $s = \frac{1}{2}$, $\psi^- = \psi^+$, and there is slow decay of correlations.

Duality!