

o) (Planar) Statistical Mechanics

6.1) Motivation

We want to describe the (macroscopic) behaviors of a system composed of a huge number of (microscopic) particles. Theoretically, if the behavior of each particle can be described, then one should be able to say something about the system. However, the number of particles is astronomical ($1 \text{ mole} \approx 10^{23}$) and it is costly (even impossible) to solve the system.

Boltzmann (end 19th century) gives a probabilistic formalism.

Ω = sample space (discrete, countable, possibly infinite)

$H : \Omega \rightarrow \mathbb{R}$ energy function (Hamiltonian)

$\beta = \frac{1}{T}$ inverse temperature

$\Rightarrow$ Boltzmann distribution $P_\beta(\omega) = \frac{1}{Z_\beta} \exp(-\beta H(\omega))$

where $Z_\beta = \sum_{\omega \in \Omega} \exp(-\beta H(\omega))$ is called
partition function.

Interpretations :

- *) At a fixed temperature, an exponential weight is associated to each configuration.
- *) At high temperature ($\beta \rightarrow 0^+$), the weights are almost a constant function of configurations, so each state is equiprobable.

*) At low temperature ($\beta \rightarrow +\infty$), the weights concentrate on minimizers of the energy function H .

$\Rightarrow$ This corresponds to the intuitions we have in physics (thermodynamics?)

Rmk: 1) $Z_\beta < \infty$ if Ω is finite, so P_β is well defined.

2) If Ω is infinite, then Z_β is not always finite. The finiteness of Z_β may depend on β .

0.2) First model (polymer / self-avoiding walk)

Consider the square lattice $\mathbb{Z}^2$ (or any other regular lattice)

Polymer = self-avoiding walk from the origin

SAW = a path that does not intersect its trajectory.

(Paul Flory, chemist, 1953)

Given a SAW ω , write $H(\omega) = |\omega| = \text{length of the walk}$.

$$\text{Then, } Z_\beta = \sum_{\omega} \exp(-\beta |\omega|) = \sum_{N \geq 1} \lambda_N \exp(-\beta N)$$

where λ_N denotes the number of SAWs of length N .

Here, the sample space $\Omega = \{\text{all SAWs}\}$ is countably infinite.

To make the model well defined, we need to ensure that $Z_\beta < \infty$.

Clearly, for $\beta=0$, $Z_\beta = \infty$. We need to understand the asymptotic behavior of λ_N (exp. growth) so that we can find β for which Z_β is finite.

Observation: for any $n, m \geq 1$, if ω is a SAW of length $n+m$, then $(\omega_1, \dots, \omega_n)$ and $(\omega_{n+1}, \dots, \omega_{n+m})$ are SAWs of length n and m .

The subadditive lemma (exercise) says that $\mu := \lim (\lambda_N)^{1/N} \in [1, \infty)$ and that $\lambda_N \geq \mu^N$ for all $N \geq 1$.

μ is called the **connective constant** (of the graph).

Let $\beta_c = \ln(\mu)$, then $Z_\beta < \infty \iff \beta > \beta_c$.

Exercise: Check the following:

1) In $\mathbb{Z}^2$, $2 < \mu < 3$

2) In $\mathbb{Z}^3$, $\mu > 3$

3) In the planar triangular lattice, $\mu > 3$.

4) In the planar hexagonal lattice, $\mu < 2$

5) For $\mathbb{Z} \times \{0, 1\}$, $\mu = \frac{1}{2}(1 + \sqrt{5})$.

Rmk:

1) It is highly non trivial to compute the connective constant of a graph. The only non trivial graph for which this value can be computed is for the planar hexagonal lattice where $\mu = \sqrt{2+\sqrt{2}}$.

(H. Duminil-Copin, S. Smirnov, 2010)

2) Although μ depends on the lattice we consider, it is conjectured that

$$Z_{\beta} = (\beta - \beta_c)^{-\gamma + o(1)} \quad \text{when } \beta \rightarrow \beta_c^+ \text{ for some } \gamma > 0$$

that only depends on the dimension of the lattice.

For example, when $d=2$, it is conjectured that $\gamma = \frac{43}{32}$.

3) Heuristically, if $\lambda_N \sim \mu^N \cdot N^{\alpha}$ then $Z_{\beta} \sim (\beta - \beta_c)^{-(1+\alpha)}$.

Thus, (conjecture) $\lambda_N = \mu^N \cdot N^{1/32 + o(1)}$.

Exercise: How to sample such a SAW? Think of a Markov process whose invariant measure is P_β ($\beta > \beta_c$).

0.3) Bernoulli percolation

Consider a graph with finite / countably many vertices and assume it has bounded degrees. Write $G = (V, E)$. (Ex: $\mathbb{Z}^2$)

Bond percolation: Given $p \in [0, 1]$. Consider a family of i.i.d. r.v.s $(w(e))_{e \in E}$ s.t. $w(e) \sim \text{Ber}(p)$.

If $w(e) = 1$, we say that e is open (we draw it on the board).

If $w(e) = 0$, we say that e is closed (we don't draw it).

We get a random subgraph from w which is given by open edges.

By abuse of notation, we still write w for the subgraph.

Site percolation: The same definition with the difference that the r.v.s are given on vertices.

Notation: $\mathbb{P}_p$ = the Bernoulli percolation measure with parameter p
 ω_p = a realization of $\mathbb{P}_p$ / $\omega_p \sim \mathbb{P}_p$.

Rmk: The terminology "Bernoulli percolation" is used for such a random subgraph model where edges / vertices are i.i.d. Bernoulli r.v.s. (independence!)

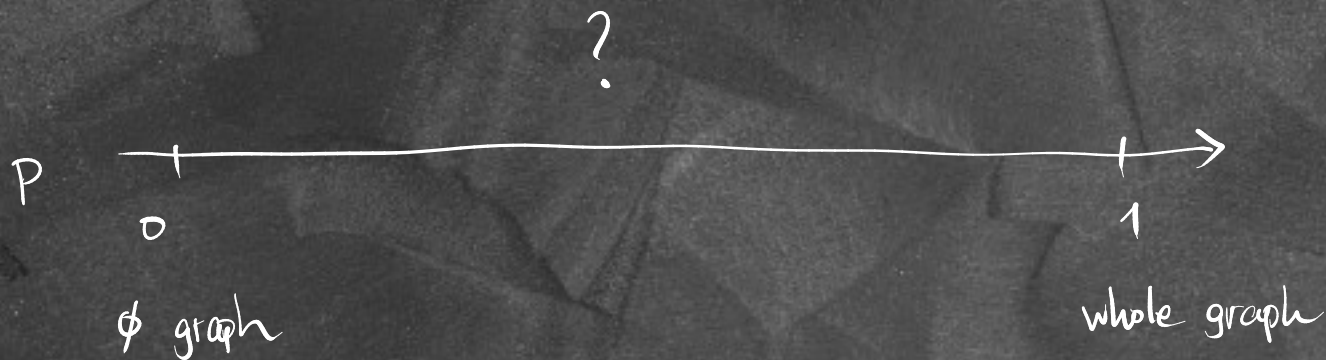
Generally speaking, without independence, we may still call it a percolation model, e.g. the random-cluster model we will see towards the end of this series of lectures.

Rmk: The independence between edges makes the "infinite-volume" measure more straightforward and we are not bothered by the "well-definedness" or the "uniqueness" of the infinite-volume measure(s).

Exercise: Show that a bond percolation is equivalent to a site percolation (on a modified graph). How about its converse? Construct some examples.

Question: Understand the behavior of the random subgraph ω when p varies.

For example, is there an infinite cluster? Is it unique? How big are the clusters if they are finite? Etc.



Exercise: Let Π_d be the d -regular tree, i.e. an infinite connected graph with countably many vertices without cycles where all the vertices are of degree d . Ex: $\Pi_2 = \mathbb{Z}$, $\Pi_3 =$  .

Exercise: Define the random-cluster model and discuss the following:

- 1) Why it can only be defined on a finite graph?
- 2) The heuristic behavior of the model when q varies.

($q \rightarrow 0^+$ UST, $q=1$ Bernoulli percolation,

$q > 1$ configurations with more clusters are preferred

$q < 1$ configurations with less clusters are preferred)

- 3) The definition of the model is "global", i.e. one needs to count the number of clusters so needs to know where a configuration looks like in the whole graph.
- 4) For the planar random-cluster model, the FK-loop representation makes the definition "local". (This may be hard to discuss at the moment, maybe I can mention it later in the lecture.)