

I. PROBABILITY THEORY

3 main ingredients:

- ▶ Ω : "sample space", "set of possible outcomes"
- ▶ \mathcal{F} : a collection of "events" $E \subset \Omega$
- ▶ \mathbb{P} : probability measure $\mathbb{P}: \mathcal{F} \rightarrow [0,1]$

I.1 FINITE PROBABILITY SPACES

Ω finite set

list the probabilities of outcomes $(p_\omega)_{\omega \in \Omega}$

$$0 \leq p_\omega \leq 1 \quad \forall \omega \in \Omega \quad (*)$$

$$\sum_{\omega \in \Omega} p_\omega = 1 \quad (**)$$

all $E \subset \Omega$ are events: $\mathcal{F} = \mathcal{P}(\Omega) = \{\text{subsets of } \Omega\}$

probability $\mathbb{P}[E] = \sum_{\omega \in E} p_\omega$

Many statistical physics models are

1. first defined on a finite probab. sp. Ω_N ($N = \text{"size parameter"}$)
2. then analyzed in the "thermodynamical limit" $N \rightarrow \infty$

Example Ising model (without external magnetic field, in a d-dim. box)

d-dimensional box $V = \llbracket 1, L \rrbracket^d \subset \mathbb{Z}^d$

\uparrow
 $\llbracket 1, L \rrbracket = \{1, 2, \dots, L\}$

$$\Omega = \{-1, +1\}^V = \{\text{functions } V \rightarrow \{-1, +1\}\}$$

$$= \{\omega = (\omega_v)_{v \in V} \mid \omega_v \in \{-1, +1\} \quad \forall v \in V\}, \quad \# \Omega = \underbrace{2^{\#V}}_{\text{finite}} = 2^{L^d}$$

set of neighbor pairs, i.e. "edges"

$$E = \{\langle v, v' \rangle \mid v, v' \in V, \|v - v'\| = 1\}$$

set of disagreeing neighbors of a config. $\omega = (\omega_v)_{v \in V}$

\leftarrow unordered pair of points, $\langle v, v' \rangle = \langle v', v \rangle$

$$E_{\text{dis}}(\omega) = \{\langle v, v' \rangle \in E \mid \omega_v \neq \omega_{v'}\}$$

probabilities $p_\omega = \frac{1}{Z} \times \#E_{\text{dis}}(\omega)$

$$\left(Z = \sum_{\omega} \#E_{\text{dis}}(\omega) \Rightarrow (***) \right)$$

Example Ehrenfest urn $N \in \mathbb{N}$, $q \in [0, 1]$

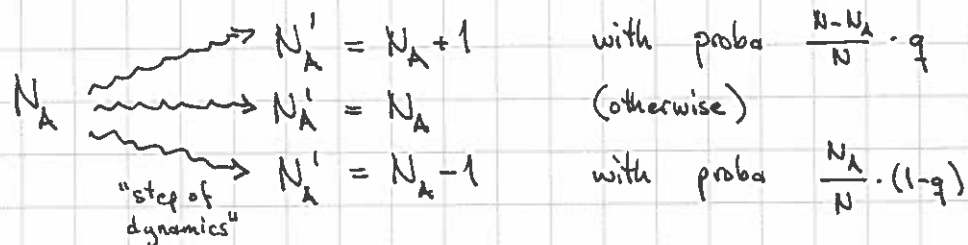
A toy model of gas in a container consisting of two parts, A and B.

N "particles", N_A in part A, $N - N_A$ in part B

Dynamics (one step, to be repeated)

- choose a particle uniformly at random

- with proba q place it in part A, otherwise in part B



Stationary distribution: N_A and N'_A have the same distribution iff $N_A \sim \text{Bin}(N, q)$

(Exercise 2/week 1)

Deterministic thermodynamical limit

"density" $\rho_A = \frac{N_A}{N}$

mean $E[\rho_A] = q$, variance $E[\rho_A] = \frac{q(1-q)}{N} \xrightarrow{N \rightarrow \infty} 0$

Uniform probability measure on a finite set

$$\#\Omega < \infty \quad \Rightarrow \quad p_\omega = \frac{1}{\#\Omega} \quad \text{where } \Omega \text{ satisfies } \textcircled{k} \text{ and } \textcircled{60}$$

trivial looking case, but often important (and can be complicated)

Example The self-avoiding polymer ~~is a~~

~~self-avoiding~~ d -dimensional (hyper)cubic lattice \mathbb{Z}^d

$$\Omega = \left\{ \text{injective } N \text{ step nearest neighbor paths} \right. \\ \left. \text{on } \mathbb{Z}^d, \text{ starting from the origin} \right\}$$

$$= \left\{ \omega = (w_j)_{j=0}^N \mid w_0 = \underline{0}, \quad \|w_j - w_{j-1}\| = 1 \quad \forall j=1, \dots, N \right. \\ \left. w_j \neq w_k \quad \forall j \neq k \right\}$$

Clearly Ω is finite, ~~the~~ $c_N := \#\Omega \leq (2d)^N < \infty$

The uniform measure $p_\omega = \frac{1}{\#\Omega} = \frac{1}{c_N}$ is a good model of the random shape of a d -dimensional polymer.

Example Random permutation $n \in \mathbb{N}$

denote ~~the~~ $[1, n] = \{1, 2, \dots, n\}$

$$S_n = \left\{ \sigma: [1, n] \rightarrow [1, n] \text{ bijective} \right\}$$

"symmetric group"
"permutation group"

~~the~~ $\#\ S_n = n!$

Take $\Omega = S_n$, $p_\omega = \frac{1}{n!}$ "uniform random permutation"

▶ a shuffled deck of cards (with $n=52$)

▶ natural model of ~~the~~ (random) input for a sorting algorithm

▶ example questions:

• ~~the~~ $P[\exists j \in [1, n] \text{ s.t. } \sigma(j) = j]$ "rencontre problem"

• the distribution of the length of the cycle that contains 1?

• probability that 1 and 2 are in the same cycle?

• expected number of cycles?

• length of the longest increasing subsequence? "last passage percolation" "nucleic growth"

1.2 KOLMOGOROV AXIOMS AND GENERAL THEORY OF PROBABILITY

- Ω a set "sample space"
- \mathcal{F} σ -algebra on Ω : a collection of subsets of Ω s.t.

(σ -1): $\emptyset \in \mathcal{F}$

(σ -2): $E \in \mathcal{F} \Rightarrow \Omega \setminus E \in \mathcal{F}$

(σ -3): $(E_n)_{n \in \mathbb{N}}, E_n \in \mathcal{F} \Rightarrow \bigcup_{n \in \mathbb{N}} E_n \in \mathcal{F}$

- \mathbb{P} probability measure on Ω : $\mathbb{P}: \mathcal{F} \rightarrow [0,1]$ such that

(\mathbb{P} -1): $\mathbb{P}[\Omega] = 1$

(\mathbb{P} -2): $E_n \in \mathcal{F}$ (for $n \in \mathbb{N}$) disjoint: $E_n \cap E_m = \emptyset \forall n \neq m$
 $\Rightarrow \mathbb{P}\left[\bigcup_{n \in \mathbb{N}} E_n\right] = \sum_{n \in \mathbb{N}} \mathbb{P}[E_n]$

Random variable $X: \Omega \rightarrow \mathcal{X}$

s.t. for all measurable $A \subset \mathcal{X}$

$$\{\omega \in \Omega \mid X(\omega) \in A\}$$

is an event

i.e. $X^{-1}(A) \in \mathcal{F}$

"measurable space", i.e. a set equipped with a σ -algebra of measurable sets

often we take $\mathcal{X} = \mathbb{R}$ with the Borel σ -algebra: the smallest σ -algebra that contains all open sets.

Expected value (integral with respect to the measure \mathbb{P})

$$X: \Omega \rightarrow \mathbb{R} \rightsquigarrow \mathbb{E}[X] = \int X d\mathbb{P} = \int_{\Omega} X(\omega) d\mathbb{P}(\omega)$$

construction:

1. indicator: $\mathbb{1}_E(\omega) = \begin{cases} 1 & \text{if } \omega \in E \\ 0 & \text{if } \omega \notin E \end{cases}, \mathbb{E}[\mathbb{1}_E] = \mathbb{P}[E]$

2. elementary random var. $X = \sum_{j=1}^n x_j \cdot \mathbb{1}_{E_j}, \mathbb{E}[X] = \sum_{j=1}^n x_j \mathbb{P}[E_j]$

3. positive random var. $X \geq 0$, approximate from below by elementary

4. real valued integrable random var. $X \in L^1(\mathbb{P})$, i.e., $\mathbb{E}[|X|] < \infty$: split to positive and negative parts

Example Finite probability spaces

Ω finite set, $\mathcal{F} = \mathcal{P}(\Omega) = \{E \subset \Omega\}$, $\mathbb{P}[E] = \sum_{\omega \in E} p_\omega$

Example Probability on integers \mathbb{Z} (or \mathbb{N})

$\Omega = \mathbb{Z}$ (or $\Omega = \mathbb{N}$)

$\mathcal{F} = \mathcal{P}(\mathbb{Z}) = \{E \subset \mathbb{Z}\}$ still ok, because Ω countable

$\mathbb{P}[E] = \sum_{n \in E} p_n$ $(p_n)_{n \in \mathbb{Z}}$, $p_n \geq 0$, $\sum_{n \in \mathbb{Z}} p_n = 1$

► Poisson (λ) : $p_n = \frac{1}{n!} \lambda^n e^{-\lambda}$ $n \in \mathbb{Z}_{\geq 0}$ ($p_n = 0 \forall n < 0$)

► Geom(r) : $p_n = r^n (1-r)$ $n \in \mathbb{Z}_{\geq 0}$ (—|—)

~~\mathbb{P}~~ $f: \mathbb{Z} \rightarrow \mathbb{R}$, $E[f] = \sum_{n \in \mathbb{Z}} p_n \cdot f(n)$

Example Probability on the real line \mathbb{R}

$\Omega = \mathbb{R}$ (or a subset of \mathbb{R})

$\mathcal{F} = \mathcal{B} = \{\text{Borel subsets of } \mathbb{R}\}$

= smallest σ -algebra that contains all open sets

\mathbb{P} for example

► Gaussian $N(\mu, \sigma^2)$: $\mathbb{P}[E] = \int_E \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2}(x-\mu)^2} dx$

► Unif($[a, b]$) : $\mathbb{P}[E] = \frac{|E \cap [a, b]|}{b-a}$ where $|\cdot|$ = Lebesgue measure

► Exp(λ) : $\mathbb{P}[E] = \int_{E \cap [0, \infty)} \lambda \cdot e^{-\lambda x} dx$

► delta measure at x_0 : $\mathbb{P} = \delta_{x_0}$ $\mathbb{P}[E] = \begin{cases} 1 & \text{if } x_0 \in E \\ 0 & \text{if } x_0 \notin E \end{cases}$

If $\mathbb{P}[E] = \int_E p(x) dx$ with some density p , then

the expected value takes the form $E[f] = \int_{\mathbb{R}} p(x) f(x) dx$

Example Brownian motion (see Part III of the course)

$\Omega = C([0, T]) = \{\text{continuous functions } [0, T] \rightarrow \mathbb{R}\}$

$\mathcal{F} = \text{Borel } \sigma\text{-algebra of the metric space } (C([0, T]), \|\cdot\|_\infty)$

\mathbb{P} Wiener measure (see Part III)

sup-norm

Example The Maxwell-Boltzmann classical ideal gas

"container" $V = [0, L] \times [0, L] \times [0, L] \subset \mathbb{R}^3$

N "particles" with "positions" $x_j \in V$ $j=1, 2, \dots, N$
 and "velocities" $v_j \in \mathbb{R}^3$ $j=1, 2, \dots, N$

$$\Omega = \underbrace{V \times V \times \dots \times V}_{N \text{ times}} \times \underbrace{\mathbb{R}^3 \times \mathbb{R}^3 \times \dots \times \mathbb{R}^3}_{N \text{ times}} \subset (\mathbb{R}^3)^{2N}$$

$\mathcal{F} = \{ \text{Borel sets} \}$ ~~scribble~~ note: $\Omega \subset \mathbb{R}^{6N}$

\mathbb{P} has a density ~~scribble~~ $(\mathbb{R}^3)^{2N} \rightarrow [0, \infty)$

$$\mathbb{P}[x_1 \in A_1, \dots, x_N \in A_N, v_1 \in U_1, \dots, v_N \in U_N] \\ = \frac{|A_1|}{|V|} \frac{|A_2|}{|V|} \dots \frac{|A_N|}{|V|} \prod_{j=1}^N \left(\left(\frac{\beta}{2\pi} \right)^{3/2} \int_{U_j} d\underline{u} e^{-\frac{\beta}{2} \|\underline{u}\|^2} \right)$$

~~scribble~~ "factorized" \leftrightarrow "all coordinates and velocities independent"

position $x_j \sim \text{Uniform}(V) = \frac{1}{|V|} \cdot \text{Lebesgue}|_V$

velocity $v_j \sim N(0, \frac{1}{\beta} \mathbb{1}_{3 \times 3})$ 3-dim. Gaussian

Let us consider a random variable on this probability space.

Let $S = \|v_1\|$ be the speed of the 1st particle, (absolute value of velocity)

What is the distribution of S ?

$$\mathbb{P}[s_{\min} \leq S \leq s_{\max}] = \mathbb{E} \left[\mathbb{1}_{\{S \in [s_{\min}, s_{\max}]\}} \right]$$

$$= \left(\frac{\beta}{2\pi} \right)^{3/2} \int_{\mathbb{R}^3} d\underline{u} e^{-\frac{\beta}{2} \|\underline{u}\|^2} \mathbb{1}_{\{\|\underline{u}\| \in [s_{\min}, s_{\max}]\}} \leftarrow \begin{cases} \text{integrated out} \\ \text{all the other} \\ \text{independent variables} \end{cases}$$

$$= \left(\frac{\beta}{2\pi} \right)^{3/2} \int_{s_{\min}}^{s_{\max}} ds 4\pi s^2 e^{-\frac{\beta}{2} s^2} \leftarrow \begin{cases} \text{changed variables} \\ \text{to radial coordinates} \end{cases}$$

$\circ \circ$ The distribution of S has density $\sqrt{\frac{2}{\pi}} \beta^{3/2} s^2 e^{-\frac{\beta}{2} s^2}$.