

Review of StatMech

Goal: Z partition f^n

Ω — region

H_Ω — Hamiltonian

$V(\Omega)$ — vol.

$S(\Omega)$ — surface area

L length
d dimension
 $V(\Omega) \sim L^d$
 $S(\Omega) \sim L^{d-1}$

B.C. (1) Periodic B.C.
(2) Hard Walls

* Fluid Continuum
* Magnet Lattice

coupling const.

$$-\frac{H_\Omega}{k_B T} = \sum_n K_n \theta_n$$

local operator

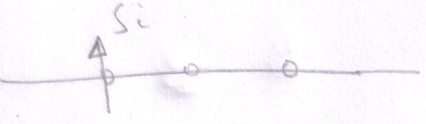
eg. $H_\Omega = \sum_{i=1}^N \left[\frac{p_i^2}{2m} + U_1(x_i) \right] + \frac{1}{2} \sum_{i \neq j} U_2(x_i, x_j) + \frac{1}{3!} \sum_{i \neq j \neq k} U_3(x_i, x_j, x_k)$

$K_1 = H, \theta_1 = \sum_i S_i$
 $K_2 = J, \theta_2 = \sum_{i,j} S_i S_j$

I SINGH MODEL

$$-\frac{H_\Omega}{k_B T} = H \sum_i S_i + J \sum_{\langle i,j \rangle} S_i S_j$$

$$Tr = \sum_{S_1} \sum_{S_2} \dots \sum_{S_{N(\Omega)}}$$



$$H \sum_i S_i$$

When Ω is finite,
no phase transition.

Partition f^n

$$Z[\{K_n\}] = Tr e^{-\beta H_\Omega}$$

$$F_\Omega[\{K\}] = -k_B T \ln Z$$

$$\beta = \frac{1}{k_B T}$$

For finite systems,

$$F_\Omega \propto V(\Omega)$$

$$\frac{\partial F_\Omega}{\partial K_n}, \frac{\partial^2 F_\Omega}{\partial K_n \partial K_m}, \dots$$

For finite system

$$F_\Omega = V(\Omega) f_b + S(\Omega) f_s + O(L^{d-2})$$

bulk free energy density

surface free energy per unit area

$$f_b[k] = \lim_{V(\Omega) \rightarrow \infty} \frac{F_\Omega}{V(\Omega)}$$

or $\lim_{N(\Omega) \rightarrow \infty} \frac{F_\Omega}{N(\Omega)}$

it exists & independent of Ω .

$$f_S[k] = \lim_{S(\Omega) \rightarrow \infty} \left\{ \frac{F_R[k] - V(\Omega) f_b[k]}{S(\Omega)} \right\}$$

$$V(\Omega) \rightarrow \infty$$

$$N(\Omega) \rightarrow \infty$$

$$v = \frac{N}{V}$$

δ

thermodynamic δ of a charged system

$\rho \leftarrow$ uniformly charged density in 3 Dimension

$$U(x) = \frac{A}{x} \sim \text{coulomb}$$

coulomb's law | at $T=0$,

Energy

$$E(R) = \int_0^R \left(\frac{4}{3} \pi r^3 \rho \right) \frac{A}{r} (4 \pi r^2 \rho dr)$$

$$E(R) = A \frac{(4\pi)^2 \rho^2 R^5}{15}$$

$$f_b = \frac{E(R)}{V(R)} = \frac{A 4\pi}{5} \rho^2 R^2 \rightarrow \text{diverges as } R \rightarrow \infty$$

$$f = E - TS^0$$

THERMODYNAMICS

0th law: Transitivity of equilibrium

1st law: $\delta Q + \delta W = dE$

Force constants: (infinitesimal) ratio of displacement to force

e.g. Isothermal compressibility $\kappa_T = -\frac{1}{V} \left. \frac{\partial V}{\partial P} \right|_T$

susceptibility of magnet $\chi_T = \frac{1}{V} \left. \frac{\partial M}{\partial B} \right|_T$

2nd law:

Revised: Kelvin's & Clausius's Statement
($Q_c = 0$) ($W = 0$)

Carnot Engine:

Entropy: $\oint \frac{\delta Q}{T} \leq 0$; $\Delta S = \int_A^B \frac{\delta Q_{rev}}{T}$; $dS \geq \frac{\delta Q}{T}$

Thermodynamic Potentials:

$$H = E - J \cdot x$$

$$F = E - TS$$

$$G = E - TS - J \cdot x$$

$$\psi = E - TS - \mu \cdot N$$

$\delta W = 0$	$\delta Q = 0$	const
const J	$\delta S > 0$	$\delta F \leq 0$
	$\delta H \leq 0$	$\delta G \leq 0$

Useful Math Results: if extensivity holds, viz $E(\lambda S, \lambda x, \lambda N) = \lambda E(S, x, N)$ then

1) $E = TS + J \cdot x + \mu \cdot N \equiv$ Fundamental eq of thermody

2) $SdT + x \cdot dJ + N \cdot d\mu = 0 \equiv$ Gibbs-Duhem relation

Probability / Sheldon Ross

summary

Binomial Theorem

$$(x+y)^n = \sum_{k=0}^n \binom{n}{k} x^k y^{n-k}$$

Chapter #2

§ 2.2 | Any subset E of the sample space is an event.

Commutative

$$E \cup F = F \cup E$$

$$EF = FE$$

Associative

$$(E \cup F) \cup G = E \cup (F \cup G)$$

$$(EF)G = E(FG)$$

Distrib

$$(E \cup F)G = EG \cup FG$$

$$E(F \cup G) = (EF) \cup (EG)$$

DeMorgan's laws:

$$\left(\bigcup_{i=1}^n E_i \right)^c = \bigcap_{i=1}^n E_i^c$$

$$\left(\bigcap_{i=1}^n E_i \right)^c = \bigcup_{i=1}^n E_i^c$$

Axioms of Probability:

$$1. \quad 0 \leq P(E) \leq 1$$

$$2. \quad P(S) = 1$$

3. For any seq. of mutually exclusive events E_1, E_2, \dots (i.e. $E_i E_j = \emptyset$ for $i \neq j$)

$$P\left(\bigcup_{i=1}^{\infty} E_i\right) = \sum_{i=1}^{\infty} P(E_i)$$

Propositions

$$1. \quad P(E^c) = 1 - P(E)$$

$$2. \quad \text{If } E \subset F, \text{ then } P(E) \leq P(F)$$

$$3. \quad P(E \cup F) = P(E) + P(F) - P(EF)$$

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Proposition:

If $\{E_n, n \geq 1\}$ is either an increasing or decreasing seq. of events, then

$$\lim_{n \rightarrow \infty} P(E_n) = P\left(\lim_{n \rightarrow \infty} E_n\right)$$

Chapter #3

Defⁿ:

If $P(F) > 0$ then

$$P(E|F) = \frac{P(EF)}{P(F)}$$

Multiplication Rule:

$$P(E_1 E_2 \dots E_n) = P(E_1) P(E_2|E_1) P(E_3|E_1 E_2) \dots P(E_n|E_1 E_2 \dots E_{n-1})$$

Baye's Formula

$$P(E) = P(EF) + P(EF^c) = P(E|F)P(F) + P(E|F^c)(1 - P(F))$$

Defⁿ:

$$\text{Odds of event: } \frac{P(A)}{P(A^c)} = \frac{P(A)}{1 - P(A)}$$

$$\text{Prop: } P(F_j|E) = \frac{P(EF_j)}{P(E)} = \frac{P(E|F_j)P(F_j)}{\sum_{i=1}^n P(E|F_i)P(F_i)} \quad \text{where } F_i \text{ are mutually exclusive s.t. } \bigcup_{i=1}^n F_i = S$$

Defⁿ:

Events E & F are independent if $P(EF) = P(E)P(F)$

Prop:

If E & F are independent, then so are E & F^c

If E & F are not independent, then they're dependent

Defⁿ:

3 events E, F, G are independent if

$$P(EFG) = P(E)P(F)P(G)$$

$$P(EF) = P(E)P(F)$$

$$P(EG) = P(E)P(G)$$

$$P(FG) = P(F)P(G)$$

Prop.

$P(\cdot|F)$ is a probability

$$a) \quad 0 \leq P(E|F) \leq 1$$

$$b) \quad P(S|F) = 1$$

c) If E_i are mutually exclusive then

$$P\left(\bigcup_{i=1}^{\infty} E_i | F\right) = \sum_{i=1}^{\infty} P(E_i | F)$$

Chapter #4

Random Variable: A real-valued f^n defined on the sample space

Discrete Random Variable:

$$p(a) = P(\{X=a\}) = \text{probability mass of } a$$

$$F(x) = P(\{X \leq x\}) = \text{cumulative distribution of } f^n \text{ of } X$$

Defⁿ:
Expectation value of X

$$E[X] = \sum_{x: p(x) > 0} x p(x)$$

Prop: $E[g(X)] = \sum g(x_i) P(x_i)$

Defⁿ:

If X is a random variable with mean μ , then the variance of X is

$$\text{var}(X) = E[(X - \mu)^2]$$

Prop:

$$\text{var}(X) = E[X^2] - (E[X])^2$$

Defⁿ:

$$\text{Standard Deviation}(X) = \sqrt{\text{Var}(X)}$$

Binomial

$$p(i) = \binom{n}{i} p^i (1-p)^{n-i} \quad | \quad E[X] = np \quad \text{var}(X) = np(1-p)$$

Poisson

$$p(i) = \frac{e^{-\lambda} \lambda^i}{i!} \quad | \quad E[X] = \text{var}(X) = \lambda$$

Geometric

$$p(i) = p(1-p)^{i-1} \quad | \quad E[X] = \frac{1}{p} \quad \text{var}(X) = \frac{1-p}{p^2}$$

Chapter #5 | CONTINUOUS RANDOM VARIABLES

$$P\{X \in B\} = \int_B f(x) dx \quad \text{where } f(x) = \text{probability density } f$$

$$F(a) = \int_{-\infty}^a f(x) dx$$

$$E[X] = \int_{-\infty}^{\infty} x f(x) dx$$

Prop: $E[g(X)] = \int_{-\infty}^{\infty} g(x) f(x) dx$ (tricky to prove)

Corollary: $E[aX+b] = aE[X] + b$

old results

$$\text{var}(X) = E[(X - \mu)^2]$$

$$= E[X^2] - (E[X])^2$$

$$\text{var}(aX+b) = a^2 \text{var}(X)$$

Chapter #6 | JOINTLY DISTRIBUTED RANDOM VARIABLES

$F(a, b) \equiv P\{X \leq a, Y \leq b\} \equiv$ Joint cumulative probability distribution

$$F_X(a) \equiv P\{X \leq a\} = P\{X \leq a, Y < \infty\} = P\left(\lim_{b \rightarrow \infty} \{X \leq a, Y \leq b\}\right)$$

$$= \lim_{b \rightarrow \infty} P\{X \leq a, Y \leq b\}$$

$$= \lim_{b \rightarrow \infty} F(a, b) = F(a, \infty)$$

$F_Y(b) \equiv F(\infty, b)$ | $F_X, F_Y \equiv$ Marginal Distributions

When X & Y are discrete,

$p(x, y) = P\{X=x, Y=y\} \equiv$ joint probability mass fⁿ.

$p_X(x) = P\{X=x\} = \sum_{y: p(x,y) > 0} p(x, y)$ | $p_Y(y) = P\{Y=y\} = \sum_{x: p(x,y) > 0} p(x, y)$

§6.2 Independent Random Variables (Earlier we talked about independent events)

Defⁿ: X & Y are independent if $P\{X \in A, Y \in B\} = P\{X \in A\} P\{Y \in B\}$

- (some work) viz. X & Y are independent if $\forall A, B$, the events $E_A = \{X \in A\}$ & $E_B = \{Y \in B\}$ are independent. these follow

$\Rightarrow P\{X \leq a, Y \leq b\} = P\{X \leq a\} P\{Y \leq b\}$

$\Rightarrow F(a, b) = F_X(a) F_Y(b)$

\nexists discrete $p(a, b) = p_X(x) p_Y(y)$ | \nexists continuous $f(x, y) = f_X(x) f_Y(y)$

Defⁿ: X & Y are jointly continuous if $\exists f(x, y) (\neq 0)$ s.t. for every set C of pairs of real numbers,

$P\{(X, Y) \in C\} = \iint_{(x,y) \in C} f(x, y) dx dy$

joint probability density fⁿ.

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for more insight $f_X(x) = \int_{-\infty}^{\infty} f(x, y) dy$ | $f_Y(y) = \int_{-\infty}^{\infty} f(x, y) dx$ } follows.

Prop: The continuous (dis) random var. X & Y are independent if & only if their joint probability density (mass) fⁿ can be expressed as

$f_{X,Y}(x, y) = h(x) g(y)$