

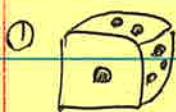
Take picture of board

01/11/2025

Probability

1. Definitions.

9:30 - 9:55



$\Omega = \{1, 2, 3, \dots, 6\}$: sample space.

$E = \{13, 12, 43, \dots\}$: event.



$\Omega = \{r \mid 0 \leq r \leq r_{\max}\}$: sample space

$E = \{r^{(1)} = 0.23, \dots, r^{(1)} = 0.1, r^{(2)} = 0.2\}$: event.

⇒ Probability : Event / Sample space.

1) Objective probabilities

$$P = \lim_{N \rightarrow \infty} \frac{N_E}{N}$$

2) Subjective probabilities.

Theoretical estimate $\sim 1/6$ dice.

Q) Monty Hall problem.

1	2	3	Result (stay #1)	Result (change to offered)
G	G	C	G	C
G	C	G	G	C
C	G	G	C	G. $\downarrow 2/3$

Rule : Host must,

- 1) open door not selected by contestant.
- 2) open door to reveal goat not car.
- 3) offer chance to switch b/w original and remaining closed.

2. Q)

9:35 - 9:40

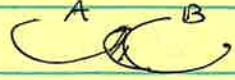
① : 20 people at least two have same birthday

$$P = 1 - \frac{365 P_{20}}{365^{20}} \approx 0.4114.$$

3. • Rules. (in HW)

9:40-9:45

- Additive : $P(A \cup B) = P(A) + P(B) - P(A \cap B)$



- Conditional probability

$$P(B|A) = \frac{P(A \cap B)}{P(A)}, \text{ Given } A.$$

(Why? Multiplicative! $\Rightarrow \underbrace{P(A)}_{A \text{ already happened}} \cdot \underbrace{P(B|A)}_{\text{Then, both } A, B \text{ happens.}} = \underbrace{P(A \cap B)}$)

- Independence.

$$P(B|A) = P(B) : \text{ doesn't care } A \text{ happens.}$$

$$\Rightarrow P(A) \cdot P(B) = P(A \cap B).$$

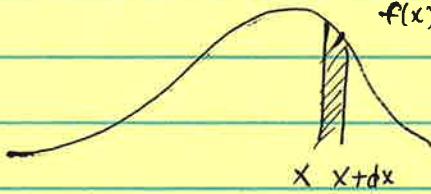
• Random Variables.

• Discrete RV

Event : $\{X = x\}$

$$\langle X \rangle = \sum_x x P(X = x)$$

• Continuous RV



$$\text{Var}(x) = \langle (X - \langle X \rangle)^2 \rangle = \langle X^2 + \langle X \rangle^2 - 2 \langle X \rangle \cdot X \rangle$$

$$f(x) \cdot dx = P\{x \leq X \leq x+dx\}.$$

$$= \langle X^2 \rangle + \langle X \rangle^2 - 2 \langle X \rangle^2$$

$$= \langle X^2 \rangle - \langle X \rangle^2$$

$$E\{X\} = \int_a^b x f(x) dx.$$

$$\sigma(x) = \{\text{Var}(x)\}^{1/2}$$

$$E\{(X - E\{X\})^2\} = \text{Var}\{X\}.$$

$\langle X^k \rangle$ = k^{th} moment.

$$= \int_a^b x^2 f(x) dx - \left(\int_a^b x f(x) dx \right)^2.$$

$\langle g(X) \rangle$ = average,

4. • Q) Prove $\langle XY \rangle = \langle X \rangle \langle Y \rangle$ if X, Y are independent

$$\langle XY \rangle = \iint y x \cdot f(x) \cdot f(y) \cdot dx dy = \int y \left(\int dx \cdot x f(x) \right) f(y) = \langle X \rangle \langle Y \rangle \#$$

$$\int x \cdot p(x=x) \cdot y \cdot p(y=y) = \int_{xy} p(x=x, y=y)$$

HW 2-c

• Multi-variate prob. dist.

- $\langle aX + bY \rangle = a\langle X \rangle + b\langle Y \rangle$
- $\text{Cov}(X, Y) = \langle (X - \langle X \rangle)(Y - \langle Y \rangle) \rangle$ $\rightsquigarrow \langle \uparrow\uparrow, \uparrow\downarrow, \downarrow\downarrow \rangle$
 $\text{mean} = 0, \text{Var}(X) \quad \text{mean} = 0, \text{Var}(Y)$
 $= \langle XY \rangle - \langle X \rangle \langle Y \rangle \cdot 2 + \langle X \rangle \langle Y \rangle = \langle XY \rangle - \langle X \rangle \langle Y \rangle$
- correlation: $\rho(X, Y) = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y}$

Q) $X_i \sim X_N$ where $X_i \sim N(\mu, \sigma^2)$, i.i.d.

Define $\bar{X} = \frac{1}{N} \sum_{i=1}^N X_i$, $\langle \bar{X} \rangle = \mu$, $\text{Var}(\bar{X}) = \sigma^2/N$, $\sigma(\bar{X}) = \sigma/\sqrt{N}$

$$\begin{aligned} \text{Pf) } \text{Var}(\bar{X}) &= \langle (\bar{X} - \langle \bar{X} \rangle)^2 \rangle = \langle \bar{X}^2 \rangle - \langle \bar{X} \rangle^2 \\ &= \frac{1}{N^2} \cdot \langle \sum_i X_i^2 \rangle = \frac{1}{N^2} \left\langle X_1^2 + X_2^2 + \dots + X_N^2 + 2X_1X_2 + \dots + 2X_{N-1}X_N \right\rangle - \mu^2 \\ &= \left(\frac{1}{N} \sum_i X_i \right) (N + 1) \cdot \langle X \cdot N \rangle + \dots + \langle X_N \cdot X \cdot N \rangle - \mu^2 \quad \frac{2}{N^2} \mu^2 \cdot \frac{N(N-1)}{2} \\ &= \left(\frac{1}{N} \sum_i X_i \right) = \frac{1}{N^2} \sum_i \langle X_i^2 \rangle + \underbrace{\frac{2}{N^2} \sum_{i \neq j} \langle X_i \cdot X_j \rangle}_{\neq 0} - \mu^2 \\ &= \frac{1}{N^2} \cdot \sum_i \{ \text{Var}(X_i) + \mu^2 \} + 0 - \mu^2 = \frac{1}{N^2} \cdot N \cdot \text{Var}(X_i) = \boxed{\sigma^2/N}. \# \end{aligned}$$

5.

Central limit Theorem \rightarrow central limit theorem.

8:50-9:55

6. • Fun: stirling's formula.

$$N! = \sqrt{2\pi N} \cdot \left(\frac{N}{e}\right)^N \rightarrow \ln N! = N \ln N - N + \dots$$

(discrete (continuous \uparrow Entropy)

If time permits.

• Gaussian distribution

$$\left(\int_{-\infty}^{+\infty} e^{-x^2} dx \right)^2 = \int_{-\infty}^{+\infty} e^{-x^2} dx \cdot \int_{-\infty}^{+\infty} e^{-y^2} dy = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} e^{-(x^2+y^2)} dx dy$$

$$x = r \cos \theta, \quad y = r \sin \theta.$$

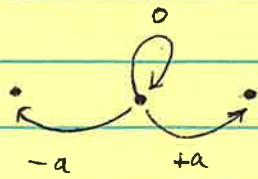
$$d\sigma = dx dy = r dr d\theta$$

$$= \int_0^{2\pi} \int_0^{+\infty} e^{-r^2} \cdot r dr d\theta = 2\pi \cdot \frac{1}{2} = (\pi)$$

$$\Rightarrow \int_{-\infty}^{+\infty} e^{-x^2} dx = \sqrt{\pi}.$$

$$\therefore f(x) = \frac{1}{\sqrt{2\pi}\sigma^2} \exp\left(-\left(\frac{(x-\mu)}{\sigma}\right)^2\right)$$

• Stationary Random Walk.



$$l_i = \begin{cases} +a & \text{prob: } p \\ 0 & \text{prob: } 1-2p \\ -a & \text{prob: } p \end{cases} \quad (p < 1/2) \quad (*)$$

Then, $\langle l_i \rangle = a p + 0 \cdot (1-2p) + (-a)p = 0 \quad \text{--- (1)}$

Also, $\langle l_i^2 \rangle = a^2 p + 0^2 \cdot (1-2p) + (-a)^2 p = 2a^2 p. \quad \text{--- (2)}$

We know that $\langle l_i l_j \rangle = \langle l_i \rangle \langle l_j \rangle = 0 \quad \text{if } i \neq j \quad \because l_i, l_j \text{ independent.}$

From the previous work,

$$\langle |x_{(n\tau)}|^2 \rangle = \langle \left(\sum_i l_i \right)^2 \rangle = \sum_i \langle l_i^2 \rangle + \sum_{i \neq j} \cancel{\langle l_i l_j \rangle} = n \cdot (2a^2 p)$$

Recall that $\langle |x(t)|^2 \rangle = 2D_s t \cdot d. \quad (d: \text{dimension} = 1.)$

$$\Rightarrow \cancel{x} \cdot (2a^2 p) = \cancel{x} D_s \cancel{n \tau} \dots$$

$$\Rightarrow D_s = \frac{a^2 p}{\tau}$$

Note that non-stationary case yields $D = \frac{a^2}{2\tau}$. and $D_s = \frac{a^2}{2\tau} (2p)$.

$$\Rightarrow D_s/D = 2p < 1 \quad (\because *) \quad \therefore D_s < D$$

Q) What is $(2p)$?

Hint: Recall $D = \cancel{k} k_B T$ or $D = k_B T / \cancel{z}$

Problem Session 1

January 10, 2025

Proof of Central Limit Theorem (CLT)

Let X_1, X_2, \dots, X_N be a random sample from an arbitrary distribution with mean μ and variance σ^2 . We shall assume that N is sufficiently large. Define,

$$\bar{X} := \frac{1}{N} \sum_{i=1}^N X_i,$$

where the expectation and the variance of \bar{X} can be calculated as,

$$\langle \bar{X} \rangle = \frac{1}{N} \sum_{i=1}^N \langle X_i \rangle = \mu, \quad \text{Var}(\bar{X}) = \langle (\bar{X} - \langle \bar{X} \rangle)^2 \rangle = \langle \bar{X}^2 \rangle - \langle \bar{X} \rangle^2 = \langle \bar{X}^2 \rangle - \mu^2,$$

Recall that $\langle \bar{X}^2 \rangle$ reads,

$$\langle \bar{X}^2 \rangle = \left\langle \left(\frac{1}{N} \sum_{i=1}^N X_i \right) \left(\frac{1}{N} \sum_{i=1}^N X_i \right) \right\rangle = \frac{1}{N^2} \left(\sum_{i=1}^N \langle X_i^2 \rangle + 2 \sum_{i \neq j} \langle X_i X_j \rangle \right),$$

and because X_i are independent samples, $\langle X_i X_j \rangle = \langle X_i \rangle \langle X_j \rangle = \mu^2$. Also note that $\langle X_i^2 \rangle = \text{Var}(X_i) + \langle X_i \rangle^2 = \sigma^2 + \mu^2$ so that,

$$\langle \bar{X}^2 \rangle = \frac{1}{N^2} \left(N (\sigma^2 + \mu^2) + 2 \binom{N}{2} \mu^2 \right) = \frac{\sigma^2 + \mu^2}{N} + \frac{N-1}{N} \mu^2 = \frac{\sigma^2}{N} + \mu^2,$$

so that the expectation and the variance of \bar{X} is,

$\langle \bar{X} \rangle = \mu, \quad \text{Var}(\bar{X}) = \frac{\sigma^2}{N}.$
--

Semi-Stationary Random Walk

Let us define a semi-stationary random walk, which is described as a random walker in 1-dimensional space. At each time step, the random walker can move either $+a$ or $-a$ with probability p and also can stay at its state with probability $1 - 2p$. We call this random walk as semi-stationary because the random walker can stay where it is.

Define a step that the random walker takes at i^{th} time step to be l_i . Then, the expectation of l_i and l_i^2 can be calculated as,

$$\langle l_i \rangle = (+a)p + (0)(1 - 2p) + (-a)p = 0, \quad \langle l_i^2 \rangle = (a^2)p + 0^2(1 - 2p) + (a^2)p = 2a^2p,$$

which can be used to calculate $\langle |X(n\tau)|^2 \rangle$ as,

$$\langle |X(n\tau)|^2 \rangle = \left\langle \left(\sum_i l_i \right)^2 \right\rangle = \sum_i \langle l_i^2 \rangle = n(2a^2p),$$

assuming that l_i and l_j are independent when $i \neq j$. Recall that $\langle |X(n\tau)|^2 \rangle = 2D_s n\tau$,

$$2D_s n\tau = n(2a^2p), \quad D_s = 2pD$$

where $D = a^2/(2\tau)$ and D_s is a diffusion coefficient for the semi-stationary random walk. Hence, we have derived that the probability of moving in the semi-stationary random walk drives the mobility of the diffusion. In other words, the action of staying at its state works as a friction which slows down the diffusion.

Lagrangian: **09:30 - 09:40**

$$F_i = m \cdot \ddot{q}_i \Leftrightarrow \frac{dp_i}{dt} = F_i \Leftrightarrow \ddot{q}_i = -\frac{1}{m} \cdot \frac{\partial V}{\partial q_i} \quad (\text{Newton, 1687})$$

Define,

$$L(q_i, \dot{q}_i) = K - V = \sum_i \left(\frac{1}{2} m \dot{q}_i^2 - V(q_i) \right) \quad \text{where,}$$

$$\frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) - \left(\frac{\partial L}{\partial q_i} \right) = 0 \quad \text{for all } i \quad (\text{Lagrange, 1760})$$

$$\text{pf: } \frac{\partial L}{\partial \dot{q}_i} = m \cdot \ddot{q}_i \equiv p_i$$

$$\frac{\partial L}{\partial q_i} = -\frac{\partial V}{\partial q_i}$$

$$\Rightarrow \frac{d}{dt} p_i = -\frac{\partial V}{\partial q_i} = F_i$$

Using derivatives,

$$dL = \sum_i \frac{\partial L}{\partial q_i} dq_i + \frac{\partial L}{\partial \dot{q}_i} d\dot{q}_i$$

$$\Rightarrow \frac{dL}{dt} = \underbrace{\sum_i \frac{\partial L}{\partial q_i} \frac{dq_i}{dt}}_{= ①} + \underbrace{\frac{\partial L}{\partial \dot{q}_i} \frac{d\dot{q}_i}{dt}}_{= \frac{\partial L}{\partial \dot{q}_i} \frac{d}{dt}(\dot{q}_i)}$$

$$\begin{aligned} ① &= \frac{\partial L}{\partial \dot{q}_i} \frac{d}{dt}(\dot{q}_i) = \left(\frac{\partial L}{\partial \dot{q}_i} \right) \cdot \left(\frac{d\dot{q}_i}{dt} \right) = \underbrace{\left(\frac{\partial \dot{q}_i}{\partial \dot{q}_i} \right)}_{= 1} \left(\frac{\partial L}{\partial \dot{q}_i} \right) \cdot \dot{q}_i \\ &= \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) \dot{q}_i \end{aligned}$$

$$\Rightarrow \frac{dL}{dt} = \sum_i \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) \dot{q}_i + \frac{\partial L}{\partial \dot{q}_i} \frac{d}{dt}(\dot{q}_i) = \sum_i \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) \dot{q}_i$$

$$\therefore \frac{d}{dt} \left[L - \sum_i \frac{\partial L}{\partial \dot{q}_i} \dot{q}_i \right] = 0$$

Hamiltonian 09:40 - 09:50

Define $H = -L + \sum_i (\partial L / \partial \dot{q}_i) \dot{q}_i \Rightarrow \frac{d}{dt}(H) = 0 \Rightarrow H \text{ is conserved.}$

Legendre's Transform (1787)

$\langle L \rightarrow H \rangle$

Note: $H = -L + \sum_i (\partial L / \partial \dot{q}_i) \dot{q}_i \text{ and } \partial L / \partial \dot{q}_i = m \cdot \ddot{q}_i \equiv p_i$

$\Rightarrow H = -L + \sum_i p_i \dot{q}_i$

Also, $\partial L / \partial \dot{q}_i = \frac{d}{dt}(\partial L / \partial \dot{q}_i) = \ddot{p}_i \rightsquigarrow \text{Force?!$

$\therefore \partial L / \partial \dot{q}_i \equiv p_i$

$\partial L / \partial q_i \equiv \dot{p}_i$

$$\begin{aligned} dH &= \sum_i p_i d\dot{q}_i + \dot{q}_i dp_i - \frac{\partial L}{\partial \dot{q}_i} d\dot{q}_i \\ &= p_i = p_i \end{aligned}$$

key point: Legendre Transform eliminates the dependence of the function L respect to the variable \dot{q}_i .

Example)

$$① H(p, x) = \frac{p^2}{2m} + V(x) : \text{Function of } p, x.$$

I want to replace "p" with something else!

$$\Rightarrow L = H - \left(\frac{\partial H}{\partial p} \right)_x \cdot p = H - V \cdot p \quad \left(\because \frac{\partial}{\partial p} \left(\frac{p^2}{2m} \right) = p/m = V \right)$$

$$\Rightarrow L = H - V \cdot p \quad : \text{constant for } (x) \text{ and } (V)$$

$$\text{check: } ① dH = \left(\frac{\partial H}{\partial p} \right)_x dp + \left(\frac{\partial H}{\partial x} \right)_p dx = (V) dp + (V'(x))_p dx \quad \boxed{\text{Function of } (p, x)}$$

$$② dL = \cancel{dp} - V dp - pdV = V dp + V'(x) dx - \cancel{dp} - pdV$$

$$= V'(x) dx - (pdV)$$

$\boxed{\text{Function of } (V, x)}$

Note: From (*),

$$dH = \sum_i -\dot{p}_i dq_i + \dot{q}_i dp_i$$

$\langle \text{Eq motion Hamilt.} \rangle$

$$\dot{p}_i = -\partial H / \partial \dot{q}_i$$

$$\dot{q}_i = \partial H / \partial p_i$$

$$dH = \sum_i \frac{\partial H}{\partial q_i} dq_i + \frac{\partial H}{\partial p_i} dp_i$$

Example on Thermodynamics.

$$E = TS - PV + \mu N \quad \text{and} \quad dE = TdS - pdV + \mu dN$$

G.T. $E(S, V, N)$

$$\textcircled{1} \quad E - \left(\frac{\partial E}{\partial S}\right)_{V, N} S = A \quad (\text{Helmholtz})$$

$$dA = dE - TdS - SdT = TdS - pdV + \mu dN - TdS - SdT \quad \boxed{N, V, T}$$

Free
Energy.

$$\textcircled{2} \quad E - TS - \left(\frac{\partial E}{\partial V}\right)_{S, N} V = G \quad (\text{Gibbs})$$

$$" - p$$

$$dG = TdS - pdV + \mu dN + pdV - TdS - SdT + Vdp \quad \boxed{N, P, T}$$

Example on pendulum. **09:40 - 09:50**

$$\textcircled{1} \quad K = \frac{1}{2} m \cdot v^2$$

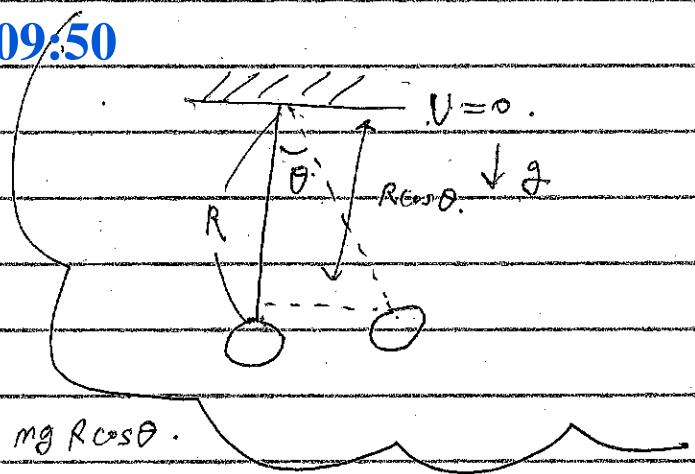
$$= \frac{1}{2} m (R \dot{\theta})^2 = \frac{1}{2} m R^2 (\dot{\theta})^2$$

$$U = - mgR \cos \theta.$$

Formulate Lagrangian as,

$$\Rightarrow \textcircled{L} = K - U = \frac{1}{2} m R^2 (\dot{\theta})^2 + mg R \cos \theta.$$

↓ function of $\theta, \dot{\theta}$



② Equation of motion.

1) Lagrangian

$$\frac{d}{dt} \left(\frac{\partial L}{\partial \dot{\theta}} \right) - \frac{\partial L}{\partial \theta} = 0$$

$$\frac{d}{dt} (m R^2 \cdot \dot{\theta}) - (-mgR \sin \theta) = 0$$

$$\Rightarrow \ddot{\theta} = -g/R \cdot \sin \theta$$

2) Newtonian

$$T = mg \cdot \cos \theta$$

$$T \cdot m \ddot{\theta} R - mgs \sin \theta = m \cdot R \ddot{\theta}$$

$$\begin{matrix} \vec{\theta} \\ mg \end{matrix}$$

$$\Rightarrow \ddot{\theta} = -g/R \cdot \sin \theta.$$

③ To Hamiltonian, (by Legendre Transform).

$L(\theta, \dot{\theta}) \rightarrow$ we want to eliminate $\dot{\theta}$

$$\begin{aligned} H^* &= -L + (\partial L / \partial \dot{\theta}) \dot{\theta} \quad \text{where } p_\theta = \partial L / \partial \dot{\theta} = mR^2 \dot{\theta} \\ &= -\frac{1}{2} mR^2 (\dot{\theta})^2 + mgR \cos \theta + mR^2 \cdot \dot{\theta} \cdot \dot{\theta} \\ &= -mgR \cos \theta + \frac{p_\theta^2}{2mR^2} \quad (\text{sign flip of } \cos) \end{aligned}$$

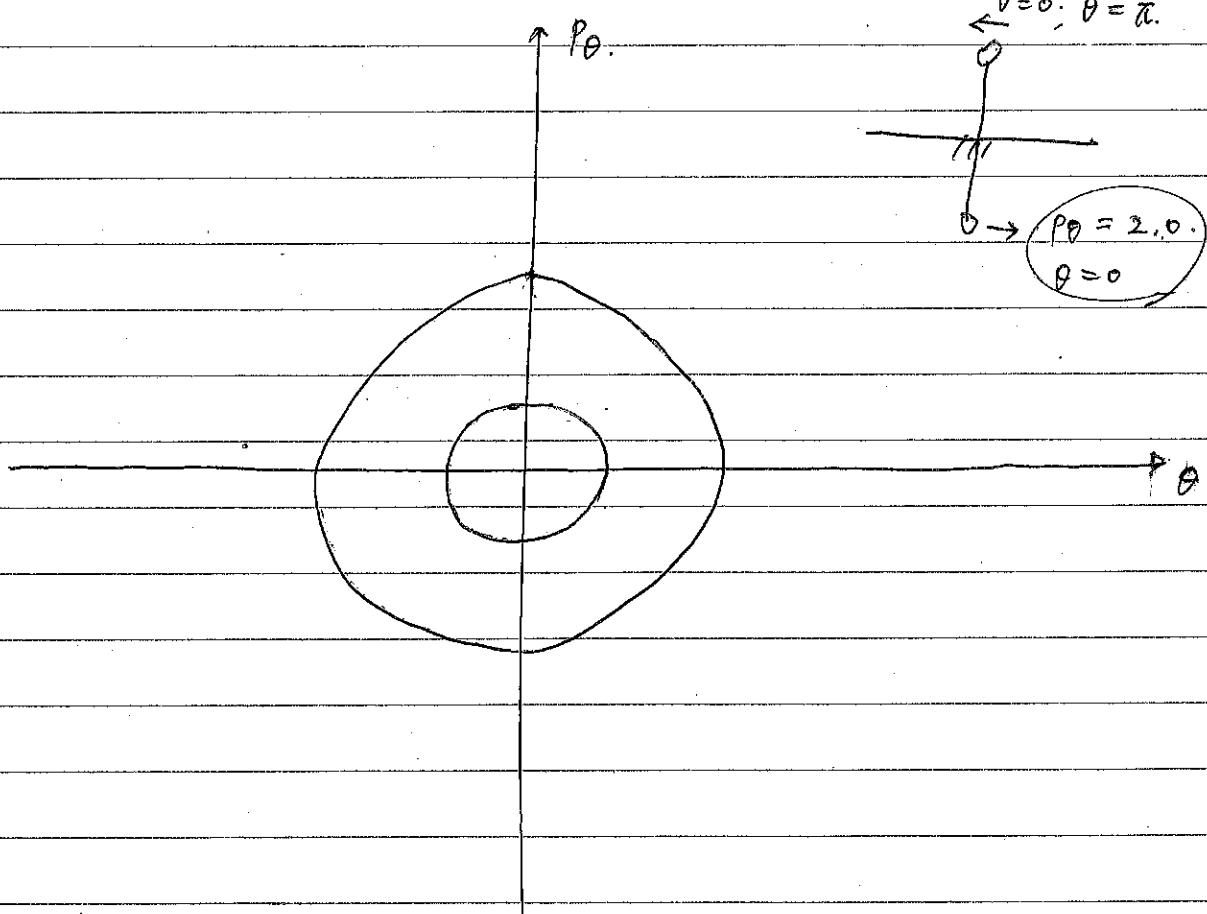
$$H^* = \frac{p_\theta^2}{2mR^2} - mgR \cos \theta$$

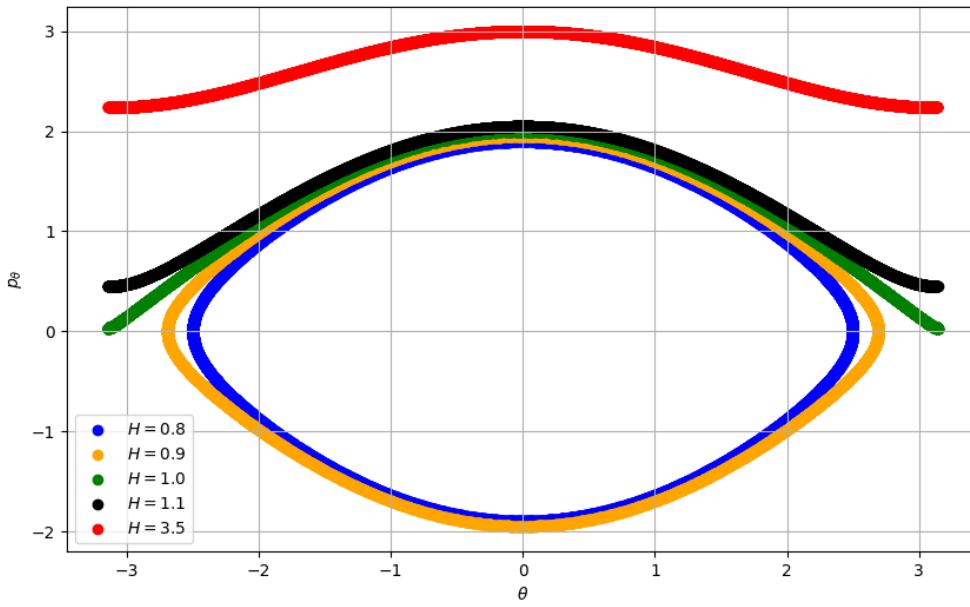
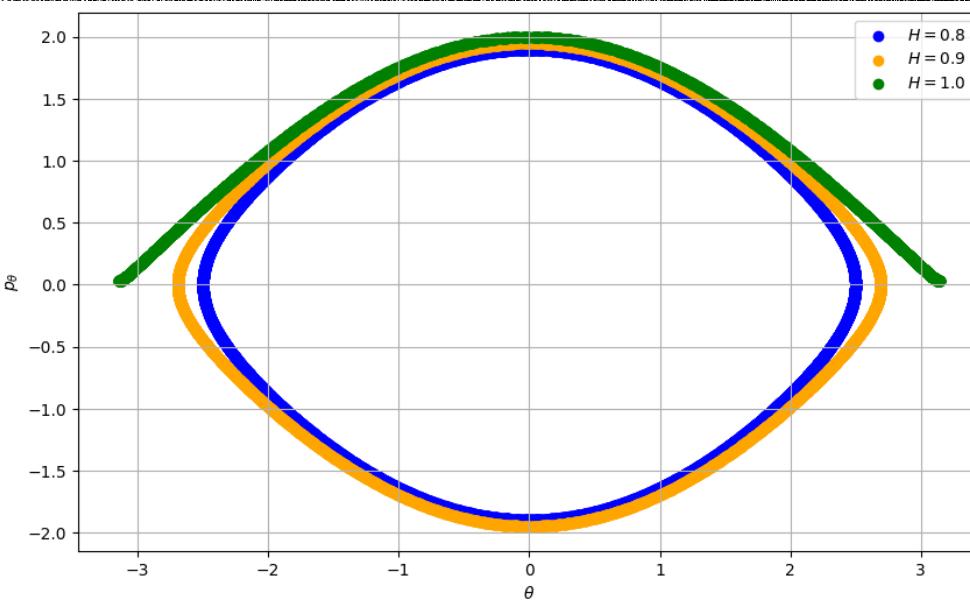
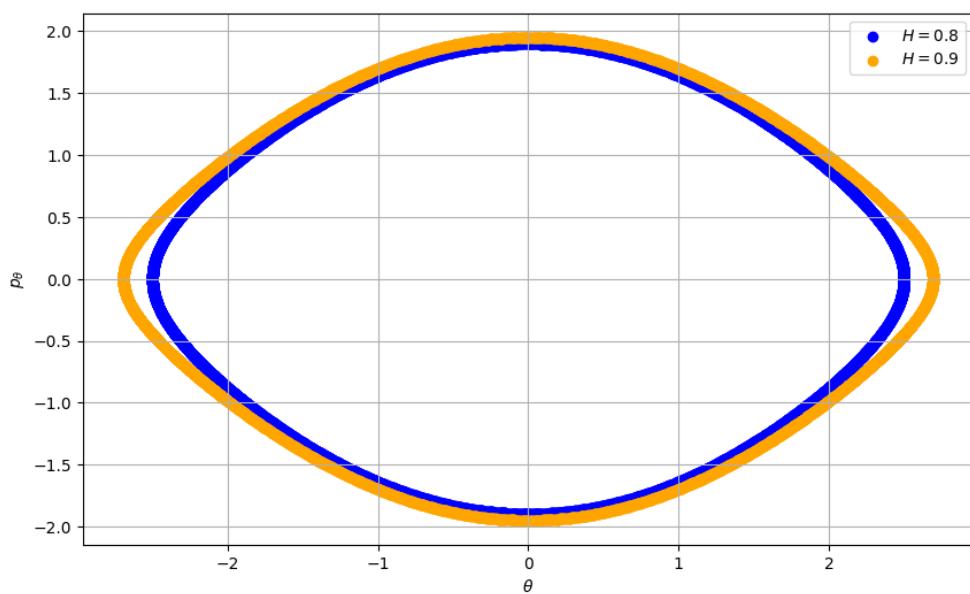
Recall eq. mot. Hamilton. **09:50 - 10:00**

$$p_i = -\partial H / \partial q_i = -(-mgR \cdot (-\sin \theta)) = -mgR \sin \theta.$$

$$q_i = +\partial H / \partial p_i = p_\theta / (mR^2)$$

$$H = 1$$





ME 346A Problem Session (01/31/2025)

- Properties of Ideal Gas. (p. 24 - 27 of notes), Ch. 6.

09:30-09:40

$$S = k_B \cdot N \left[\frac{5}{2} + \log \left(\frac{V}{N} \cdot \left(\frac{4\pi m E}{3Nk_B^2} \right)^{3/2} \right) \right] \quad \text{Refer p. 10 Ch 7}$$

for derivation with
hypersphere assumption

(a) $E(S, V, N) = ?$, $T = ?$, $p = ?$, $\mu = ?$

Does $pV = Nk_B T$ hold?

$$E(S, V, N) = \frac{3Nh^2}{4\pi m} \left(\frac{N}{V} \right)^{2/3} \exp \left[\frac{2S}{3Nk_B} - \frac{5}{3} \right]$$

$V^{-2/3}$

$$\textcircled{1} \quad T = \left(\frac{\partial E}{\partial S} \right)_{V, N} = \frac{2}{3Nk_B} \cdot E(S, V, N) \Rightarrow E = \frac{3}{2} Nk_B T.$$

$$\textcircled{2} \quad p = - \left(\frac{\partial E}{\partial V} \right)_{S, N} = - \left(- \frac{2}{3} \cdot V^{-5/3} \cdot \textcircled{1} \right) = \frac{2}{3} \cdot \frac{1}{V} E \\ = Nk_B T / V \Rightarrow pV = Nk_B T.$$

$$\textcircled{3} \quad \mu = \left(\frac{\partial E}{\partial N} \right)_{S, V} = E/N \cdot \left(\frac{5}{3} - \frac{2S}{(3Nk_B)} \right)$$

09:40-09:50

$$(b) A = E - TS = \frac{3}{2} Nk_B T - T Nk_B \left[\log \left(\frac{V}{N} \left(\frac{2\pi m k_B T}{h^2} \right)^{3/2} \right) + \frac{5}{2} \right]$$

$$\textcircled{4} \quad G = E - TS + pV = A + pV \\ = -Nk_B T \left[\log \left(\frac{p}{T} \left(\frac{2\pi m k_B T}{h^2} \right)^{3/2} \right) + \textcircled{3} \right]$$

From \textcircled{3}, pf (a),

$$\mu N = E \left(\frac{5}{3} - \frac{2S}{(3Nk_B)} \right) = \frac{5}{2} Nk_B T - TS$$

$$= \textcircled{G} \Rightarrow G = \mu N.$$

09:50-10:00

$$\text{-(d)} \quad C_V = \left(\frac{\partial Q}{\partial T} \right)_{V,N} = T \cdot \left(\frac{\partial S}{\partial T} \right)_{V,N} = -T \cdot \left(\frac{\partial^2 A}{\partial T^2} \right)_{V,N}$$

$$\text{since } S = -(\partial A / \partial T)_{N,V}$$

$$\therefore A = E - TS$$

$$\Rightarrow C_V = T \left(\frac{\partial S}{\partial T} \right) = \frac{3}{2} N k_B.$$

$$\text{-(e)} \quad C_P = \left(\frac{\partial Q}{\partial T} \right)_{P,N} = T \left(\frac{\partial S}{\partial T} \right)_{P,N} = -T \left(\frac{\partial^2 G}{\partial T^2} \right)_{P,N}$$

$$\text{since } S = -(\partial G / \partial T)_{P,N}$$

$$\therefore G = E - TS + PV$$

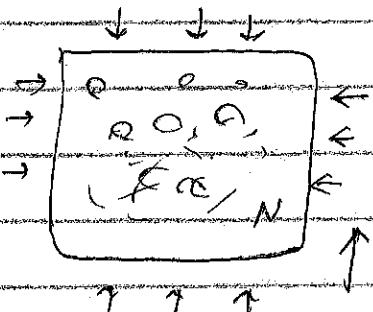
$$\Rightarrow C_P = \frac{5}{2} N k_B.$$

$$\Rightarrow C_P - C_V = N k_B.$$

10:00-10:10

(e) At constant pressure, and fixed N , we want to know

$$\alpha = \frac{1}{V} \left(\frac{\partial V}{\partial T} \right)_{P,N}$$



$$\text{From } PV = Nk_B T, V = \frac{Nk_B T}{P}.$$

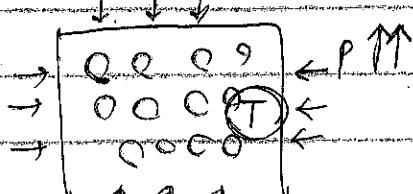
$$\Rightarrow \alpha = \frac{1}{V} \left(\frac{\partial V}{\partial T} \right)_{N,P} = \frac{1}{Nk_B T} \cdot \frac{Nk_B}{P} = 1/T$$

$$\Rightarrow \alpha = 1/T$$

At constant temperature, and fixed N , we want to know $\beta = \frac{\partial^2 V}{\partial P} = \frac{1}{T^2} V/T^2$

$$\beta = -\frac{1}{V} \left(\frac{\partial V}{\partial P} \right)_{T,N}$$

$$\beta = +\frac{1}{V} \cdot \left(\frac{Nk_B T}{P^2} \right) = 1/P$$



$$\Rightarrow \beta = 1/P$$

$$= N k_B = C_P - C_V$$

10:10-10:20

Notes on Legendre Transform & Laplace Transform.

$$Z(\beta) = \sum_{\nu | E(\nu) = E} e^{-\beta E(\nu)} \approx \int dE \cdot \Omega(E) \cdot e^{-\beta E} : \text{Laplace Transform.}$$

$$\begin{aligned} \textcircled{1} &= \int dE \cdot \Omega(E) e^{-\beta E} = \int dE \cdot \exp \left[-N(\beta E - \log w(E)) \right] \\ &\quad \approx w(E) \\ &\quad = A(E) \end{aligned}$$

$$\text{where } A = E - T \cdot S = E - (\partial E / \partial S)_{T, V} \cdot S : \text{Legendre Transform}$$

\textcircled{1} & \textcircled{2} are related.

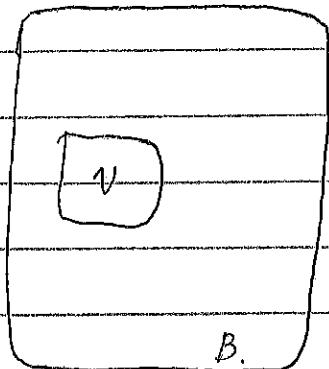
< Optional >

ME346A

Problem Session. (01/31/2025).

01/31/2025.

- Micro-canonical Ensemble.



$$E_T = E(n) + E_B \quad \rightarrow \text{partition function.}$$

$$p(V, B) = 1 / \Omega_{\text{T}}(N_T, V_T, E_T)$$

$$\Rightarrow p(n) = \sum_{\{B \mid E(n) + E_B = E_T\}} p(N_B, B) = \frac{1}{\Omega_T} \cdot \Omega_B$$

$$\Rightarrow p(n) = \frac{1}{\Omega_T} \cdot \Omega_B(E_T - E(n), N_B, V_B)$$

$N_T, V_T, E_T \rightarrow \text{Fixed.}$

$$\log p(n) \propto \log \Omega_B(E_T - E(n), N_B, V_B) = \log \Omega_B(E_T)$$

$$- \left(\frac{\partial \log \Omega_B(N_B, V_B, E_B)}{\partial E_B} \right) \cdot E(n)$$

$$+ O((E(n))^2)$$

$$\Rightarrow \log p(n) \propto - \underbrace{\left(\frac{\partial \log \Omega_B}{\partial E_B} \right)}_{N_B, V_B} E(n)$$

$(= \beta)$

$$\Rightarrow p(n) \propto e^{-\beta E(n)} \quad ; \text{Boltzmann Distribution!}$$

$$\text{~~Definition~~ } Z(\beta) = \sum_E e^{-\beta E} = \frac{1}{E} \cdot \Omega(E) \cdot e^{-\beta E} \quad <\text{Laplace}>$$

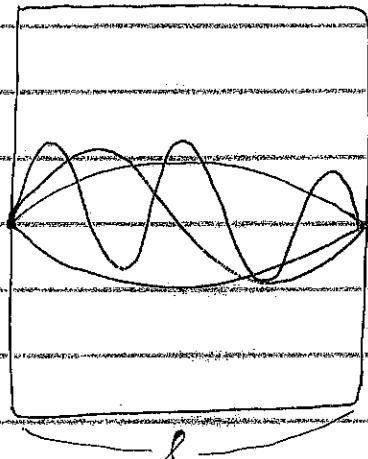
$$\Rightarrow \text{~~Definition~~} \approx \int dE \cdot \Omega(E) e^{-\beta E} = \int dE \exp(-\beta [E - \beta^{-1} \log \Omega(E)])$$

$$= \int dE \exp \left(-N\beta \cdot \underbrace{(\varepsilon - T_S)}_{= a} \right) \quad <\text{Legendre}>$$

• Ideal Gas Law.

→ Translational only,

$$\text{Quantum Mechanics} = \frac{n^2 h^2}{8\pi M l^3} = \varepsilon_n \quad (\text{avertized})$$



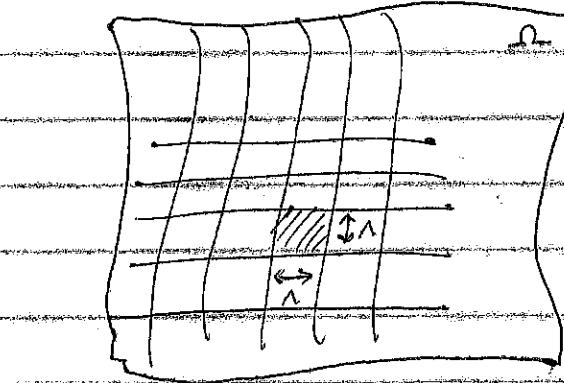
$$Z(\beta) = \sum_{n=1}^{\infty} e^{-\beta \varepsilon_n}.$$

$$Z(\beta) \approx \int_0^{\infty} e^{-\beta \varepsilon_n} dn = \frac{\sqrt{2\pi m k_B T}}{h} \quad (\beta \ll 1)$$

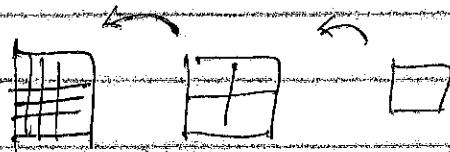
(2)

$$\text{Extend (2) to 3D, } \varepsilon_{n,3D} = \frac{h^2}{8\pi m} \left[\left(n_x/l_x \right)^2 + \left(n_y/l_y \right)^2 + \left(n_z/l_z \right)^2 \right]$$

$$\Rightarrow Z(\beta)_{3D} = \left(\frac{2\pi m k_B T}{h^2} \right)^{3/2} \cdot l_x l_y l_z = V / \lambda^3 \quad (\lambda = (\sqrt{2\pi m k_B T} / h)^{1/3})$$



λ : Thermal De Broglie wavelength.



$$A = -\beta^{-1} \log Z(\beta), \quad A = E - TS, \quad Z(\beta) = \frac{1}{N!} (Z(\beta))^N$$

$$\Rightarrow -p = \frac{\partial A}{\partial V} \Big|_{N,T} = -\beta^{-1} \cdot \frac{\partial}{\partial V} \left(\frac{1}{N!} \log \left(\frac{V}{\lambda^3} \right) \right)^N = -\beta^{-1} \cdot \frac{N}{V}$$

$$\therefore pV = N \cdot k_B T \Leftrightarrow \rho P = \rho.$$

- From Quantum Mechanics Partition Function.

$$Z(\rho) = \frac{1}{N!} \cdot z^N(\rho) \quad (\text{where } z(\rho) = V/\lambda^3)$$

$$\langle E \rangle = \frac{\partial}{\partial(-\rho)} \log Z(\rho) = Nk_B T^2 \left(\frac{\partial}{\partial T} \log z(\rho) \right)_{N,V}$$

$$= Nk_B T^2 \cdot \frac{\frac{3}{2} \cdot \left(\frac{2\pi m k_B T}{h^2} \right)^{1/2}}{\left(\frac{2\pi m k_B T}{h^2} \right)^{3/2}} = \frac{3}{2} Nk_B T^2 \cdot \frac{1/\lambda}{V/\lambda^3} \cdot \frac{1}{T}$$

$$\Rightarrow \boxed{\langle E \rangle = \frac{3}{2} Nk_B T.} \quad \text{as } N \rightarrow \infty, \quad \underline{\underline{E = \frac{3}{2} Nk_B T}} \quad (\because \text{var}(E) \downarrow 0)$$

- Some thermo-properties.

$$\rho = -(\partial E / \partial V)_{S,N} = Nk_B T / V$$

$$\mu = (\partial E / \partial N)_{S,V} = \frac{3}{2} k_B T.$$

$$\alpha = \frac{1}{V} \left(\frac{\partial V}{\partial T} \right)_P = 1/T, \quad \gamma = -\frac{1}{V} \left(\frac{\partial V}{\partial P} \right)_{T,V} = 1/\rho.$$

\Rightarrow similar to Sackur-Tetrode equation.

Problem Session 02/07/2025

09:30 → 09:40

Ideal Gas.

$$Z = \frac{1}{N!} \frac{1}{h^{3N}} \cdot \int_{i=1}^{3N} d\mathbf{q}_i d\mathbf{p}_i \exp\left(-\frac{H(\{\mathbf{q}_i\}, \{\mathbf{p}_i\})}{k_B T}\right)$$

$$H = \sum_i \frac{|\mathbf{p}_i|^2}{2m} + \cancel{U(\mathbf{q}_i)}$$

$$\Rightarrow Z = \frac{1}{N!} \frac{1}{h^{3N}} \cdot \int_{i=1}^{3N} \underbrace{d\mathbf{q}_i}_{\text{d}\mathbf{p}_i} d\mathbf{p}_i \exp\left(-\frac{P^2}{2mk_B T}\right)$$

$$= \frac{1}{N!} \frac{1}{h^{3N}} \cdot V \cdot \left[\int_{-\infty}^{\infty} dP \cdot \exp\left(-\frac{P^2}{2mk_B T}\right) \right]^{3N}$$

$$= \frac{V^N}{N! h^{3N}} (2\pi m k_B T)^{3N/2}$$

$$\therefore Z = \frac{V^N}{N! h^{3N}} (2\pi m k_B T)^{3N/2} \quad \text{--- ①}$$

$$\therefore A = -k_B T \ln Z = -k_B T \cdot N \left[\ln \left(\frac{V}{N} \left(\frac{2\pi m k_B T}{h^2} \right)^{3/2} \right) + 1 \right]. \quad \#$$

09:40 → 09:45

Quantum (if time allows).

$$\epsilon_n = \frac{n^2 h^2}{8ml^2} \Rightarrow Z = \prod_{n=1}^{\infty} e^{-\rho \epsilon_n} = \frac{\sqrt{2\pi m k_B T}}{h} \quad (\rho \ll 1).$$

$$\text{In 3D, } Z = \left(\frac{2\pi m k_B T}{h^2} \right)^{3/2} dx dy dz = V / l^3$$

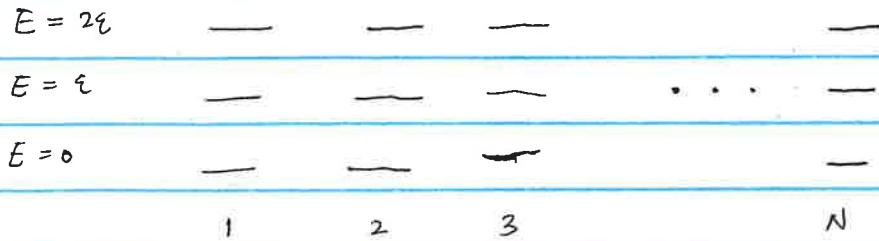
$$\text{With } N \text{ particles, } Z^N \cdot \frac{1}{N!} = \frac{V^N}{N! h^{3N}} (2\pi m k_B T)^{3N/2} \quad \text{--- ②}$$

\therefore ② recovers ①.

At high temp limit, classical \equiv quantum.

09:45 → 09:55

Molecules with 3 energy.



~~$$\mathcal{Z} = \prod_{i=1}^N e^{-\beta E_i}$$~~

$$\mathcal{Z} = \prod_{i=1}^N e^{-\beta E_i} \quad H = \sum_{i=1}^N \epsilon_i n_i$$

$$\begin{aligned} 1-1) \quad e^{-\beta H} &= \exp\left(-\beta \cdot \epsilon_1 \cdot \sum_{i=1}^N n_i\right) = \exp(-\beta \epsilon n_1) \cdot \exp(-\beta \epsilon n_2) \dots \\ &= \underbrace{\prod_{i=1}^N e^{-\beta \epsilon n_i}} \end{aligned}$$

$$\begin{aligned} 1-2) \quad \Rightarrow \mathcal{Z} &= \sum_{\{n_i\}} \prod_{i=1}^N e^{-\beta \epsilon n_i} = \prod_{i=1}^N \left(\sum_{n_i=0,1,2} e^{-\beta \epsilon n_i} \right) \\ &= \underbrace{(1 + e^{-\beta \epsilon} + e^{-2\beta \epsilon})}_{}^N \end{aligned}$$

$$\begin{aligned} 2) \quad A &= -k_B T \ln \mathcal{Z} = -N k_B T \ln (1 + e^{-\beta \epsilon} + e^{-2\beta \epsilon}) \\ S &= -\frac{\partial A}{\partial T} = -N k_B \cdot \ln (1 + e^{-\beta \epsilon} + e^{-2\beta \epsilon}) - N k_B T \cdot \frac{(-\epsilon e^{-\beta \epsilon} + (-2\epsilon) e^{-2\beta \epsilon})}{1 + e^{-\beta \epsilon} + e^{-2\beta \epsilon}} \left(-\frac{1}{k_B T^2} \right) \\ \Rightarrow S &= N k_B \cdot \ln (1 + e^{-\beta \epsilon} + e^{-2\beta \epsilon}) + N k_B T \cdot \underbrace{\frac{1}{k_B T^2}}_{\sim} \frac{e^{-\beta \epsilon} + 2e^{-2\beta \epsilon}}{1 + e^{-\beta \epsilon} + e^{-2\beta \epsilon}} \\ &= N/T \end{aligned}$$

$$\begin{aligned} E &= A + TS = N(\epsilon \cdot e^{-\beta \epsilon} + 2\epsilon e^{-2\beta \epsilon}) \\ &\quad \underbrace{\frac{1 + e^{-\beta \epsilon} + e^{-2\beta \epsilon}}{1 + e^{-\beta \epsilon} + e^{-2\beta \epsilon}}}_{\text{Expectation - 1}} \end{aligned}$$

$$09:55 \rightarrow 10:03) \quad \langle E \rangle = -\frac{\partial}{\partial \beta} \ln Z = \frac{\partial \cdot \log(Z)}{\partial(-\beta)} = \frac{\partial}{\partial(-\beta)} \cdot N \cdot \log(1 + e^{-\beta \epsilon} + e^{-2\beta \epsilon}).$$

$$= \frac{\epsilon e^{-\beta \epsilon} + 2\epsilon e^{-2\beta \epsilon}}{1 + e^{-\beta \epsilon} + e^{-2\beta \epsilon}} \cdot N.$$

\sim Expectation.

$$= N \cdot \epsilon \left(\frac{e^{-\beta \epsilon} + 2e^{-2\beta \epsilon}}{1 + e^{-\beta \epsilon} + e^{-2\beta \epsilon}} \right)$$

10:00 \rightarrow 10:05 4) Heat capacity,

$$C_V = \left(\frac{\partial E}{\partial T} \right)_N$$

$$= \frac{\partial \beta}{\partial T} \cdot \frac{\partial E}{\partial \beta} = -\frac{1}{k_B T^2} \frac{\partial E}{\partial \beta} = \frac{1}{k_B T^2} \cdot \frac{\partial E}{\partial(-\beta)}.$$

$$= \frac{1}{k_B T^2} \cdot N \cdot \epsilon \cdot \frac{\left[(e^{-\beta \epsilon} + 4\epsilon e^{-2\beta \epsilon})(e^{-\beta \epsilon} + e^{-2\beta \epsilon}) - (e^{-\beta \epsilon} + 2e^{-2\beta \epsilon})(\epsilon)(e^{-\beta \epsilon} + 2e^{-2\beta \epsilon}) \right]}{(1 + e^{-\beta \epsilon} + e^{-2\beta \epsilon})^2}$$

$$= \frac{1}{k_B T^2} \cdot N \cdot \left[\frac{\epsilon^2 e^{-\beta \epsilon} + 4\epsilon^2 e^{-2\beta \epsilon}}{1 + e^{-\beta \epsilon} + e^{-2\beta \epsilon}} - \left(\frac{\epsilon e^{-\beta \epsilon} + 2\epsilon e^{-2\beta \epsilon}}{1 + e^{-\beta \epsilon} + e^{-2\beta \epsilon}} \right)^2 \right]$$

\sim $\sqrt{\text{Var}(E)} \quad \langle \epsilon^2 \rangle \quad \langle \epsilon \rangle^2$

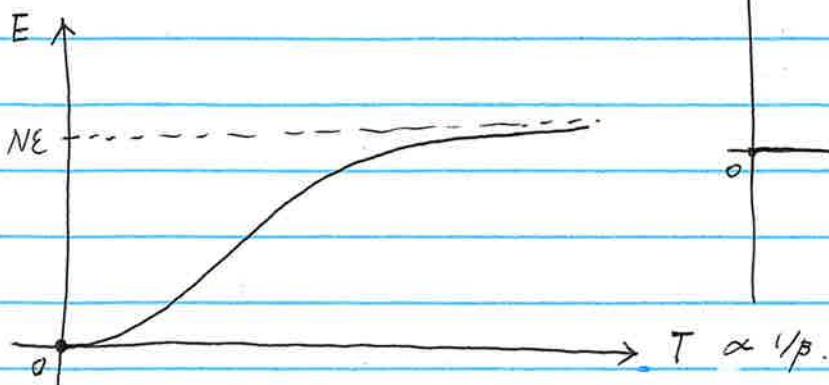
$$= \langle \epsilon^2 \rangle - \langle \epsilon \rangle^2 = \text{Var}(\epsilon)$$

$\sim //$

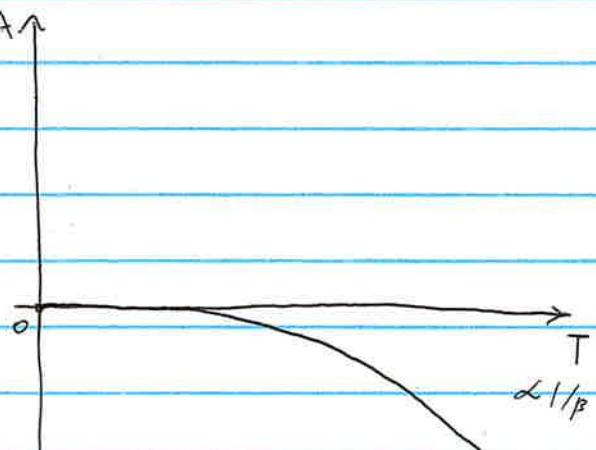
$$A = -Nk_B T \ln(1 + e^{-\beta\epsilon} + e^{-2\beta\epsilon})$$

10:05 → 10:15 5)

$$E = N \left(\frac{\epsilon e^{-\beta\epsilon} + 2\epsilon e^{-2\beta\epsilon}}{1 + e^{-\beta\epsilon} + e^{-2\beta\epsilon}} \right)$$

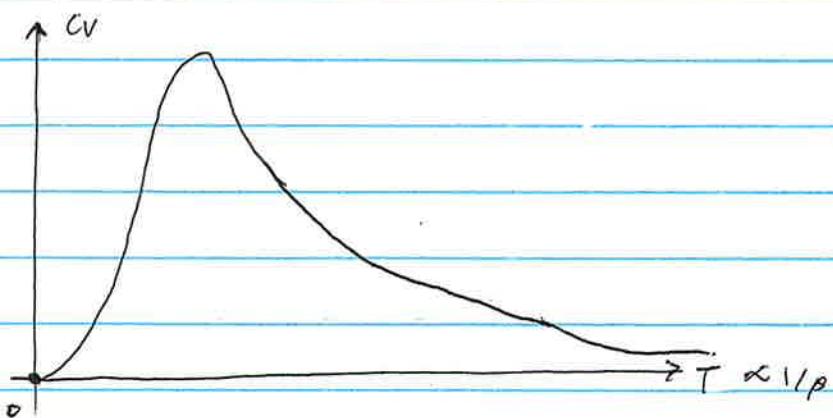
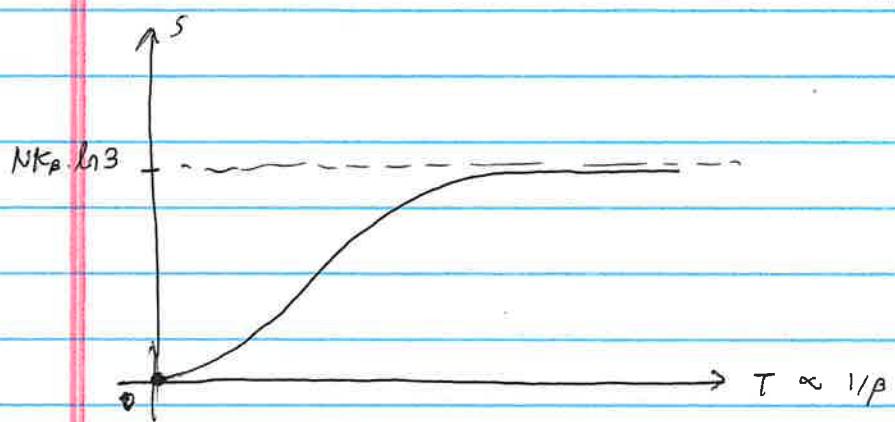


$A \uparrow$

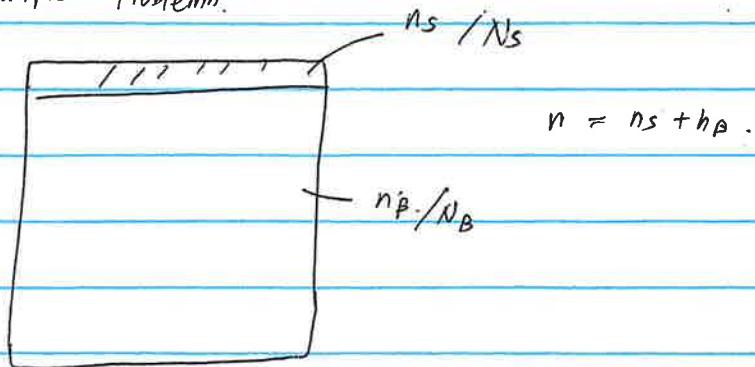


$$S = Nk_B \ln(1 + e^{-\beta\epsilon} + e^{-2\beta\epsilon}) + \frac{N}{k_B \cdot \beta} \cdot \frac{\epsilon e^{-\beta\epsilon} + 2\epsilon e^{-2\beta\epsilon}}{1 + e^{-\beta\epsilon} + e^{-2\beta\epsilon}}$$

$\underbrace{\phantom{Nk_B \ln(1 + e^{-\beta\epsilon} + e^{-2\beta\epsilon})}}_{= N/T}$



$10:15 \rightarrow 10:30$ Example Problem.



$$(a) E_s = -n_s \cdot \varepsilon$$

$$S_s = k_B \cdot \ln \left(\frac{N_s!}{n_s! (N_s - n_s)!} \right)$$

$$A_s = E_s - T \cdot S_s = -n_s \cdot \varepsilon - T \cdot k_B \cdot \ln \left(\binom{N_s}{n_s} \right)$$

$$(b) E_B = -\cancel{n_B} \varepsilon$$

$$S_B = k_B \cdot \ln \left(\frac{N_B!}{n_B! (N_B - n_B)!} \right)$$

$$A_B = E_B - T \cdot S_B = -\cancel{n_B} \varepsilon - T \cdot k_B \cdot \ln \left(\binom{N_B}{n_B} \right)$$

(c) Using sterling's formula,

$$A_s \approx -n_s \cdot \varepsilon - k_B T \cdot \left[N_s \cdot \ln N_s - n_s \cdot \ln n_s - (N_s - n_s) \ln (N_s - n_s) \right]$$

$$A_B \approx 0 - k_B T \cdot \left[N_B \cdot \ln N_B - n_B \cdot \ln n_B - (N_B - n_B) \ln (N_B - n_B) \right]$$

$$\begin{aligned} \frac{\partial A}{\partial n_s} &= -\varepsilon + k_B T \left[\ln n_s + 1 - \ln (N_s - n_s) - 1 \right] \\ A &= A_s + A_B \quad - k_B T \left[\ln n_B + 1 - \ln (N_B - n_B) - 1 \right] \quad (\because n - n_s = n_B) \end{aligned}$$

$$= -\varepsilon - k_B T \cdot \ln \left[\frac{(N_s - n_s)(n - n_s)}{n_s(N_B - n_B)} \right]$$

$$\text{Now, } N_s, N_B \gg n_s, n_B \Rightarrow \frac{\partial A}{\partial n_s} \approx -\varepsilon - k_B T \cdot \ln \left[\frac{(n - n_s)N_s}{n_s \cdot N_B} \right] = 0$$

(d) In (c), reads,

$$n_s = \frac{N_s \cdot n}{N_B \cdot \exp(-\beta \epsilon) + N_s} \quad \text{--- (*)}$$

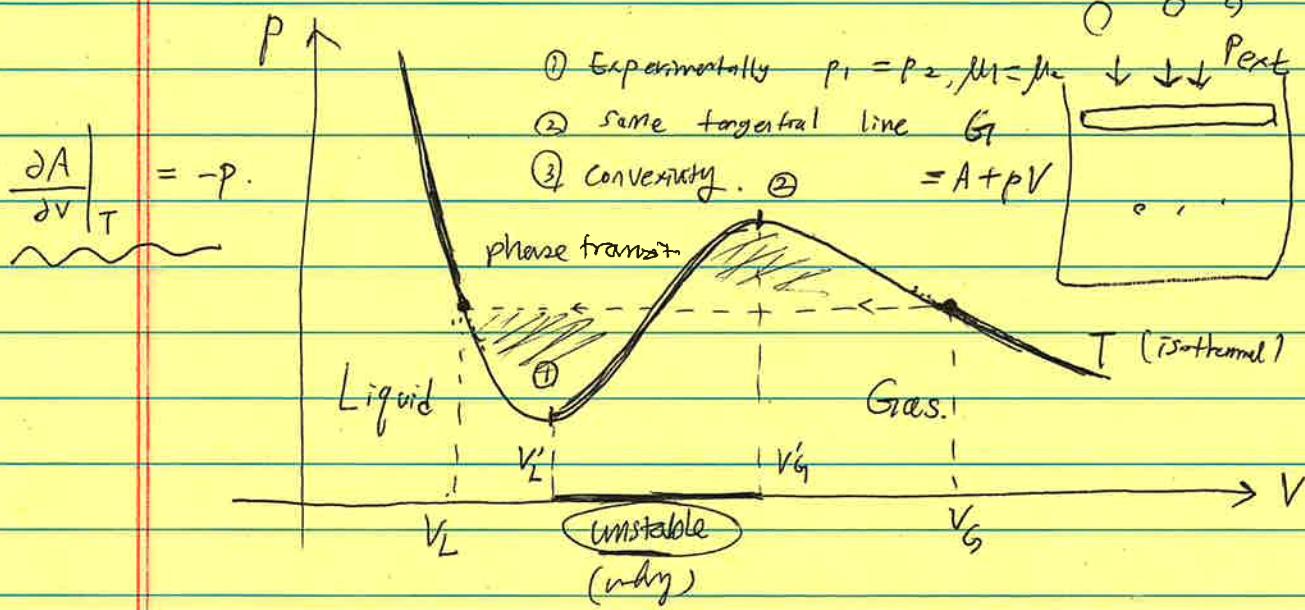
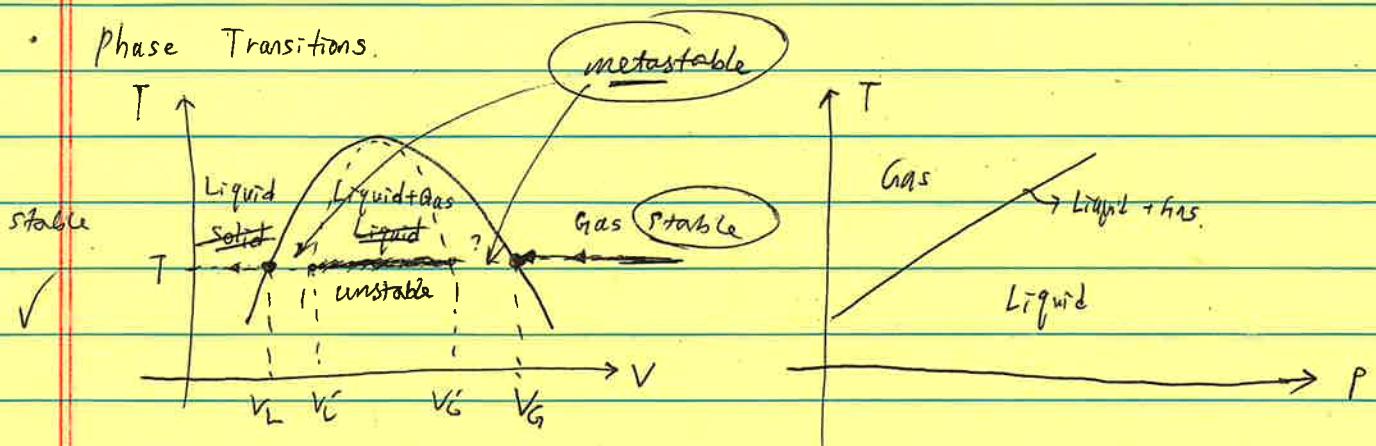
Also, $\frac{\partial A_s}{\partial n_s} = -\epsilon - k_B T \cdot \ln \left(\frac{N_s - n_s}{N_s} \right) = \mu_s$

$$\frac{\partial A_B}{\partial n_B} = k_B T \cdot \ln \left(\frac{N_B - n_B}{N_B} \right) = \mu_B.$$

$\mu_s = \mu_B$ recovers. (*) from Free energy minimization.

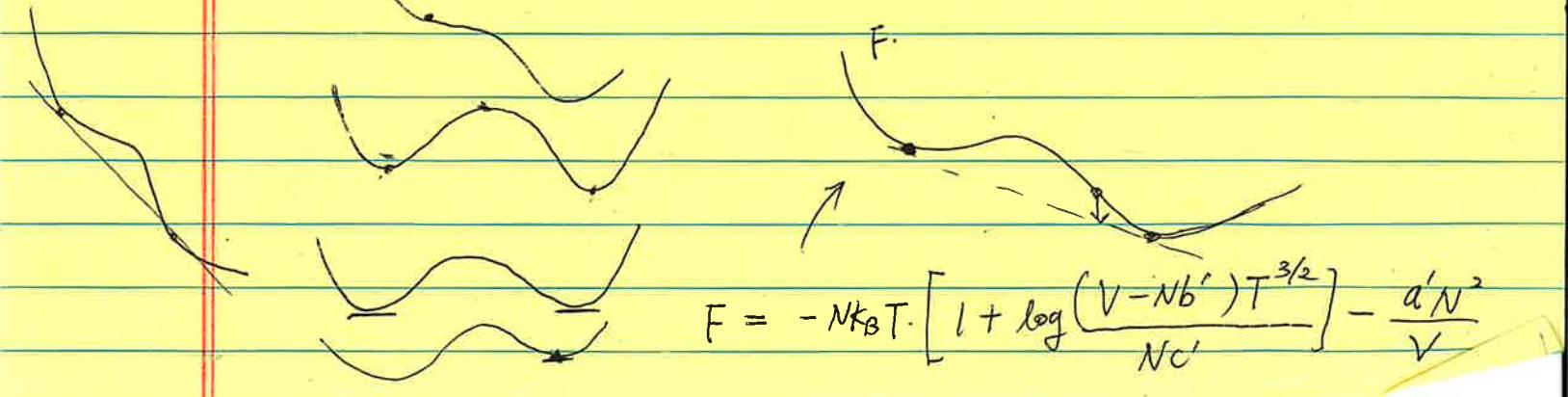
Free Energy Minimization \equiv Chemical potential equilibration, ~~if~~

• Phase Transitions.



In region ① ②, in a cylinder, decreasing P_{ext} slightly will introduce increase in volume. Such volume increase (according to graph ①②), will increase the pressure. Then, again, volume increases. Repeats \Rightarrow Unstable.

✓ \Rightarrow Maxwell's construction $\left. \frac{\partial P}{\partial V} \right|_T < 0$ (necessary for stability)



$$\beta P = \rho$$

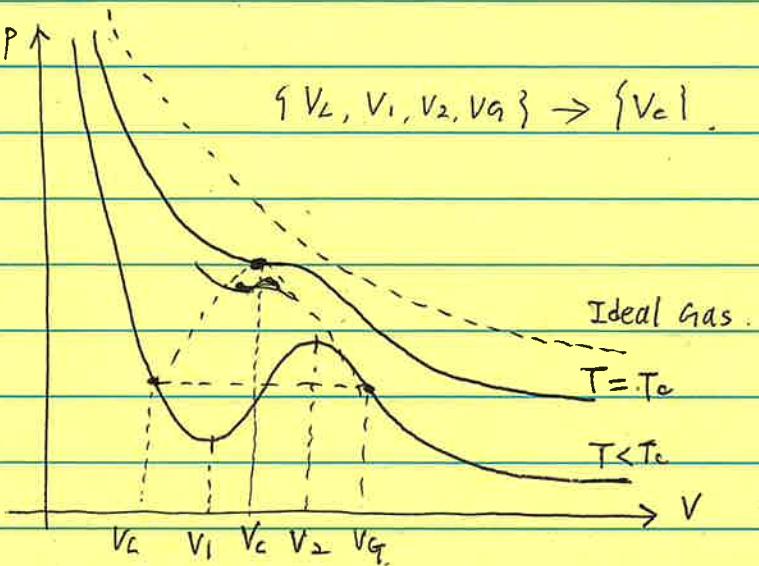
- Critical Point of Van der Waals Model.

$$P = \frac{k_B T}{V-b} - \frac{a}{V^2} \quad (*)$$

$$\{V_L, V_1, V_2, V_G\} \rightarrow \{V_c\}$$

(At $T = T_c$)

$$\begin{aligned} \textcircled{1} \frac{\partial P}{\partial V} \Big|_{V_c} &= 0 \\ \textcircled{2} \frac{\partial^2 P}{\partial V^2} \Big|_{V_c} &= 0 \end{aligned} \quad \left. \begin{array}{l} \text{Inflection} \\ \text{point!} \end{array} \right\}$$



$$\begin{aligned} \textcircled{1} \frac{\partial P}{\partial V} &= -k_B T \cdot (V_c - b)^{-2} + 2a V_c^{-3} = 0 \\ \Rightarrow 2a V_c^{-3} &= k_B T_c (V_c - b)^{-2} \end{aligned}$$

$$\begin{aligned} \textcircled{2} \frac{\partial^2 P}{\partial V^2} &= 2k_B T (V_c - b)^{-3} - 6a V_c^{-4} = 0 \\ \Rightarrow 6a V_c^{-4} &= 2k_B T_c (V_c - b)^{-3} \end{aligned}$$

$$\Rightarrow \frac{1}{3} \cdot V_c^{+1} = \frac{1}{2} \cdot (V_c - b)^{+1} \Rightarrow 3V_c - 3b = 2V_c \Rightarrow \underline{\underline{V_c = 3b}}$$

$$(*) \Rightarrow P_c = \frac{k_B T_c}{2b} - \frac{a}{9b^2}$$

$$\textcircled{1} \Rightarrow 2a (3b)^{-3} = k_B T_c \cdot (2b)^{-2} \Rightarrow \underbrace{k_B T_c}_{\sim} = \frac{8a}{27b}, \quad \underbrace{P_c}_{\sim} = \frac{a}{27b^2}$$

- Non-dimensional variables

$$\hat{P} = P/P_c, \quad \hat{V} = V/V_c, \quad \hat{T} = T/T_c.$$

$$\begin{aligned} \frac{\hat{P}}{\hat{V}^2} &= \frac{8\hat{T}/27}{3\hat{V}-1} - \frac{1}{9\hat{V}^2} \\ \Rightarrow \hat{P} \cdot P_c &= \frac{k_B \hat{T} \cdot T_c}{\hat{V} \cdot V_c - b} - \frac{a}{(\hat{V} \cdot V_c)^2} \\ \Rightarrow \hat{P} \cdot \frac{a}{27b^2} &= \frac{\frac{bK}{27b} \hat{T}}{3b\hat{V}-b} - \frac{a}{(3b\hat{V})^2} \end{aligned} \quad \Rightarrow \underbrace{\left(\hat{P} + \frac{3}{\hat{V}^2} \right) (3\hat{V}-1)}_{(*)'} = 8\hat{T}$$

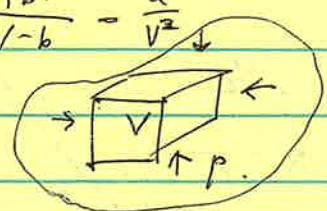
• Critical Exponent (k_T)

$$\text{Given } \left(\hat{p} + \frac{3}{\hat{V}^2}\right) \left(3\hat{V} - 1\right) = 8\hat{T}, \Leftrightarrow p = \frac{k_B T}{V-b} - \frac{a}{V^2}$$

(1) what happens $\hat{V} \approx 1/3 \Leftrightarrow V \approx \frac{1}{3} V_c$

$$(2) \text{ Isothermal compressibility } k_T = -\frac{1}{V} \frac{\partial V}{\partial p}$$

$$k_T = -\frac{1}{V} \cdot \frac{\partial V}{\partial p}.$$



$$\rightarrow \text{watch } \frac{\partial p}{\partial V} = -k_B T \cdot (V-b)^{-2} + 2a V^{-3}$$

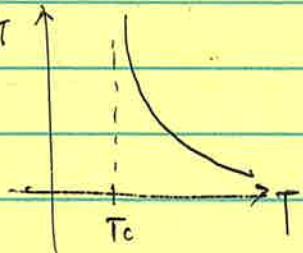
$$\rightarrow \frac{\partial V}{\partial p} = \frac{1}{-k_B T (V-b)^{-2} + 2a V^{-3}} \Rightarrow k_T = \frac{1}{k_B T \cdot V (V-b)^{-2} - 2a V^{-2}}$$

$$\rightarrow \text{Around } T = T_c, V = V_c = 3b$$

$$k_T = \frac{1}{k_B T \cdot 3b (3b)^{-2} - 2a (3b)^{-2}} \quad \left(\text{and apply } a = \frac{27}{8} b k_B T_c \right)$$

$$\Rightarrow k_T = \frac{1}{\frac{3k_B}{4b} T - \frac{3k_B}{4b} T_c} = \frac{4b}{3k_B} \frac{1}{(T - T_c)}$$

$$\boxed{\gamma = 1}$$



• Critical Exponent (ν) $T \rightarrow T_c$

(a) How does volume deviate around T_c ?

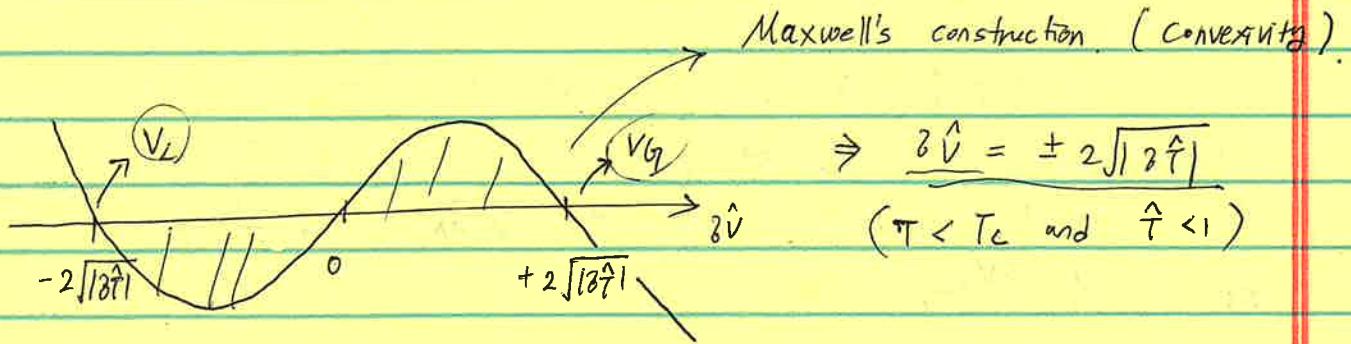
We know non-dimensional relation.

$$\hat{p} = \frac{8\hat{T}}{3\hat{V} - 1} - \frac{3}{\hat{V}^2} \quad \left\{ \begin{array}{l} \hat{T} \rightarrow 1 + \beta \hat{T} \\ \hat{V} \rightarrow 1 + \beta \hat{V} \end{array} \right\} \quad 4(1 + \beta \hat{T})$$

$$\hat{p} = \frac{8(1 + \beta \hat{T})}{2 + 3\beta \hat{V}} - \frac{3}{(1 + \beta \hat{V})^2} = \frac{4(1 + \beta \hat{T})}{1 + \frac{3}{2} \beta \hat{V}} - \frac{3}{(1 + \beta \hat{V})^2}$$

$$\approx 1 + 4\beta \hat{T}$$

$$\Rightarrow \hat{p} \propto 1 + 4\beta \hat{T} - 6(\beta \hat{T})(\beta \hat{V}) \left(1 + \frac{(\beta \hat{V})^2}{4\beta \hat{T}}\right).$$



$$\Rightarrow \hat{v} = 1 \pm 2\sqrt{1 - \frac{\tau}{\tau_c}} \Rightarrow \hat{v} = \frac{v}{v_c} = 1 \pm \sqrt{1 - T/T_c}$$

$$\Rightarrow v_g - v_l = 4v_c \left(1 - \frac{T}{T_c}\right)^{1/2} = 12b \cdot \left(1 - \frac{T}{T_c}\right)^{1/2}$$

Thns, $v_g - v_l$ goes to zero following $\propto (T_c - T)^{1/2}$
 $= (T_c - T)^{\tilde{\beta}}$

$$\tilde{\beta} = 1/2$$

- Problem Session (ME346A).
- Monte Carlo Simulation (Rosenbluth, Teller, Metropolis) at Markov chain.

03/08/2025

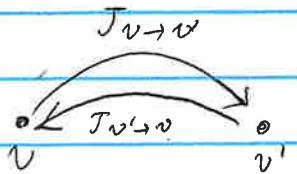
- ① Initial state v
 - ② Generate random state v'
 - ③ Evaluate probability
- $$\frac{p(v')}{p(v)} = \frac{\frac{1}{Z} \cdot e^{-\beta E(v')}}{\frac{1}{Z} \cdot e^{-\beta E(v)}} = e^{-\beta(E(v') - E(v))} = e^{-\beta \Delta E}$$
- ④ Accept / reject with probability,
- $$p_{acc}(v \rightarrow v') = \min(1, e^{-\beta \Delta E}).$$

- Detailed Balance.

$$p(v) \cdot p(v \rightarrow v') = p(v') \cdot p(v' \rightarrow v)$$

$$\Leftrightarrow \frac{p(v \rightarrow v')}{p(v' \rightarrow v)} = \frac{p(v')}{p(v)}$$

Note: $\underbrace{p(v) \cdot p(v \rightarrow v')}$ \approx "flux."



If you select any pairs of v and v' , (v, v')

$$\text{Fluxes } J_{v \rightarrow v'} = J_{v' \rightarrow v}$$

Global detailed balance \rightarrow No net flow

No net flow \rightarrow Reversible.

Reversible \rightarrow Equilibrium distribution is sampled.

- Examine detailed balance for MCMC.

$$P(v \rightarrow v') = \underbrace{P_{\text{gen}}(v \rightarrow v')}_{\substack{\text{probability to generate} \\ \text{such move}}} \cdot \underbrace{P_{\text{acc}}(v \rightarrow v')}_{\substack{\text{probability to accept} \\ \text{such move.}}}$$

From detailed balance,

$$\frac{P(v)}{P(v')} = \frac{P(v' \rightarrow v)}{P(v \rightarrow v')} = \frac{P_{\text{gen}}(v' \rightarrow v) \cdot P_{\text{acc}}(v' \rightarrow v)}{P_{\text{gen}}(v \rightarrow v') \cdot P_{\text{acc}}(v \rightarrow v')}$$

When generating (proposing) probabilities are "symmetric", (ie: $P_{\text{gen}}(v \rightarrow v') = P_{\text{gen}}(v' \rightarrow v)$)

$$\frac{P(v)}{P(v')} = \frac{P_{\text{acc}}(v' \rightarrow v)}{P_{\text{acc}}(v \rightarrow v')}$$

Metropolis-Hastings.

Apply Monte carlo, where

$$P_{\text{acc}}(v \rightarrow v') = \min [1, e^{-\beta \Delta E}] \quad (\Delta E = E(v') - E(v))$$

$$\frac{P(v' \rightarrow v)}{P(v \rightarrow v')} = \frac{P(v)}{P(v')} = \frac{\min [1, e^{+\beta \Delta E}]}{\min [1, e^{-\beta \Delta E}]} \quad \begin{aligned} &\text{should be, } \frac{e^{-\beta E(v)}}{e^{-\beta E(v')}} = e^{+\beta (E(v') - E(v))} \\ &= e^{\beta \Delta E} \end{aligned}$$

Condition for D.B.

Case 1) $\Delta E > 0$

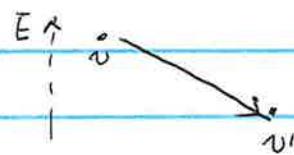


$$\min [1, e^{-\beta \Delta E}] = e^{-\beta \Delta E}$$

$$\min [1, e^{+\beta \Delta E}] = 1$$

$$P(v)/P(v') = e^{+\beta \Delta E}$$

Case 2) $\Delta E < 0$



$$\min [1, e^{-\beta \Delta E}] = 1$$

$$\min [1, e^{+\beta \Delta E}] = e^{\beta \Delta E}$$

$$P(v)/P(v') = e^{\beta \Delta E}.$$

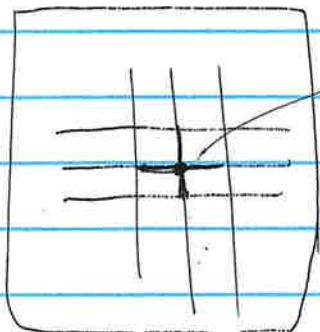
- Some notes on Monte Carlo.

① How do you implement it?

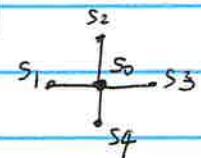
Main question: How to probability

EX) `np.random.rand() < 0.3`

② How to calculate energy difference in Ising Model?



$$H = -J \sum_{<(i,j)} s_i s_j - h \sum_i s_i$$



$$\text{Before: } -h s_0 - h s_1 - \dots - h s_4$$

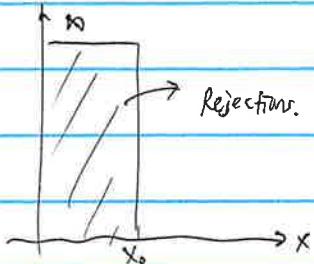
$$\text{After: } +h s_0 - h s_1 - \dots - h s_4$$

$$\text{Before: } -J s_0 s_1 - J s_0 s_3 - J s_0 s_2 - J s_0 s_4$$

$$\text{After: } -J(-s_0) s_1 - J(-s_0) s_3 - J(-s_0) s_2 - J(-s_0) s_4$$

$$\begin{aligned} \Delta H &= -h(-s_0 - s_0) - J(s_1 + \dots + s_4) \cdot (-s_0 - s_0) \\ &= 2 \left(-h \cdot s_0 - J \cdot s_0 (s_1 + \dots + s_4) \right) \end{aligned}$$

③ Hard disk model



(1) Equation of state calculations by Fast Computing Machines. (1953)

$$\cdot F_i = m \ddot{q}_i \quad (1687)$$

$$\underbrace{\frac{dp_i}{dt}}_{=F_i} = F_i \quad \text{--- ①}$$

$$\cdot L = K - U = \sum_i \frac{1}{2} m(\dot{q}_i)^2 - U(q_i)$$

$$\underbrace{\frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) - \frac{\partial L}{\partial q_i}}_{=H_i} = 0 \quad \text{--- ②}$$

$$\underbrace{\frac{d}{dt} \left(m \dot{q}_i \right)}_{\downarrow} + \left(+ \frac{\partial}{\partial \dot{q}_i} U(q_i) \right) = 0$$

$$\Rightarrow \ddot{q}_i = - \frac{1}{m} \nabla U(q_i)$$

$$\ddot{q}_i = \underbrace{- \frac{1}{m} \nabla}_{\textcircled{1}} \underbrace{\frac{\partial U}{\partial q_i}}_{\textcircled{2}} \sim ①$$

$$\frac{dL}{dt} = \sum_i \underbrace{\frac{\partial L}{\partial q_i} \frac{dq_i}{dt}}_{(1)} + \underbrace{\frac{\partial L}{\partial \dot{q}_i} \frac{d\dot{q}_i}{dt}}_{(2)} \\ = \frac{dL}{d\dot{q}_i} \frac{d}{dt} (\dot{q}_i)$$

$$(1) = \frac{\partial L}{\partial \dot{q}_i} \frac{d\dot{q}_i}{dt} = \left(\frac{\partial \dot{q}_i}{\partial \dot{q}_i} \right) \frac{\partial L}{\partial \dot{q}_i} \left(\frac{d\dot{q}_i}{dt} \right) \dot{q}_i \\ = \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) \dot{q}_i \\ \Rightarrow \underbrace{\frac{d}{dt} \left[\frac{\partial L}{\partial \dot{q}_i} \dot{q}_i \right]}_u + \underbrace{\frac{dL}{d\dot{q}_i} \frac{d}{dt} [\dot{q}_i]}_v$$

$$(uv)' = u'v + uv'$$

$$\Rightarrow \frac{dL}{dt} = \frac{d}{dt} \left[\underbrace{\frac{\partial L}{\partial \dot{q}_i}}_u \underbrace{\dot{q}_i}_v \right] \Rightarrow \underbrace{\frac{d}{dt} \left[-L + \frac{\partial L}{\partial \dot{q}_i} \dot{q}_i \right]}_H = 0$$

$$H = -L + \underbrace{\frac{\partial L}{\partial \dot{q}_i}}_{\neq p_i} \dot{q}_i = -L + p_i \dot{q}_i$$

$$\text{Ex) } H = \frac{p^2}{2m} + U(x) \quad \textcircled{H(x, p)}$$

$$\textcircled{m} \quad p = mv$$

$$L = -H + \left(\frac{\partial H}{\partial p} \right) p = -H + v \cdot p.$$

$$\textcircled{1} \quad dH = \frac{\partial H}{\partial p} dp + \frac{\partial H}{\partial x} dx = v dp + \underbrace{U'(x) dx}_{\sim}$$

$$\textcircled{2} \quad \underbrace{\frac{dL}{dt}}_{L(x, v)} = -dH + v dp + pdv = -v dp - \cancel{U'(x) dx} + \cancel{v dp} + p dv$$

$$\textcircled{L(x, v)}$$

$$\underbrace{H = -L + \left(\frac{\partial L}{\partial v} \right) v}$$

Quiz

01/17/2025

- If RVs X, Y satisfy $\langle XY \rangle = \langle X \rangle \langle Y \rangle$, they are independent.

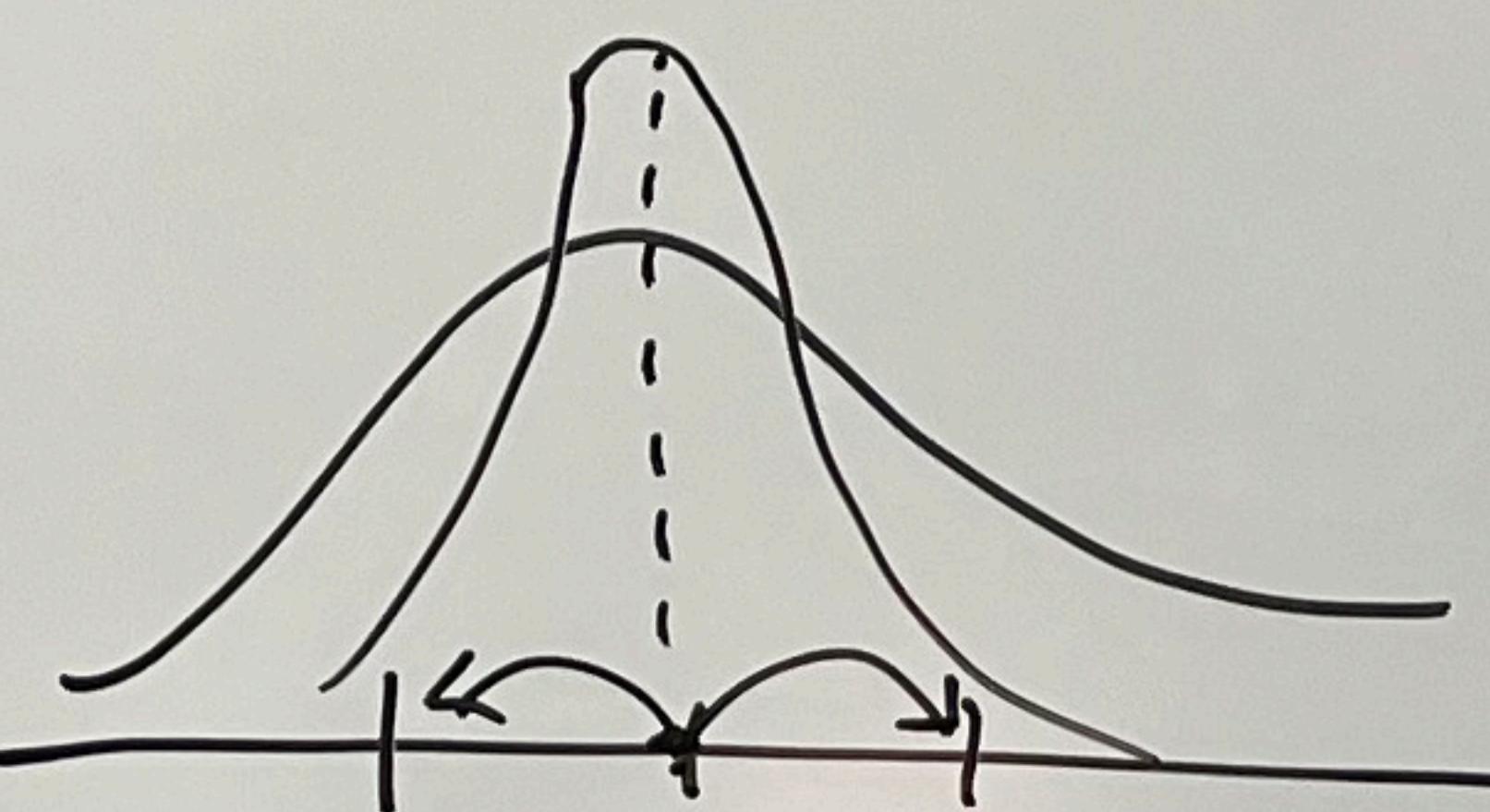
F

$$J = \frac{N_{\text{net}}}{\text{Area} \cdot \text{Time}} \quad T$$

$$\begin{array}{ll} l_i = \pm 1 & \text{prob: } p = 1/2 \\ 0 & \text{prob: } 1 - 2p \end{array}$$

F

0 0



$$Y = X^2$$

$$X = [-1, +1]$$

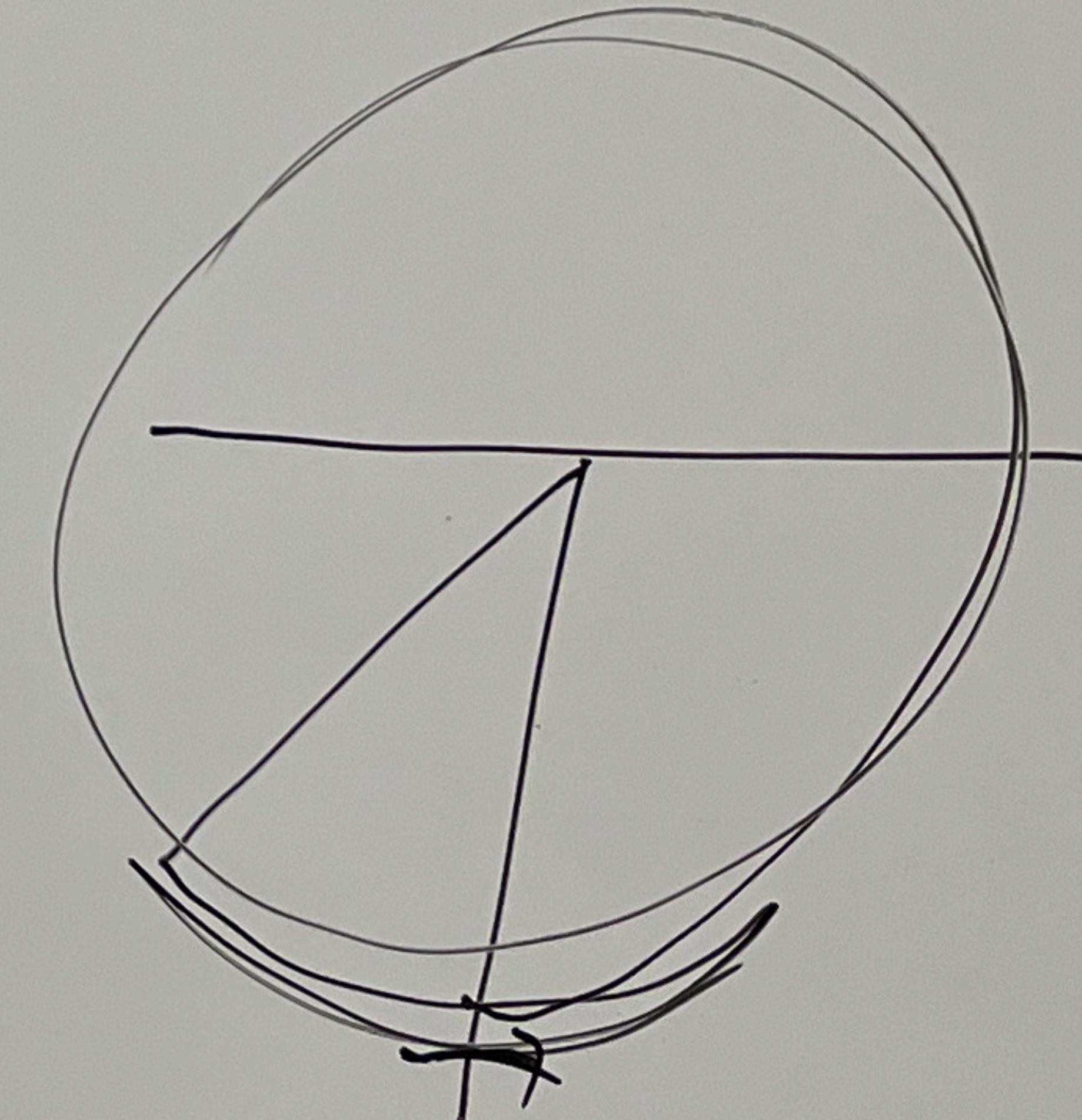
$$P(Y=y | X=x) = p(Y=y)$$

- $L = K - U$, L is conserved.

F

- U is conservative $\frac{dH}{dt} = 0 \Rightarrow$ conservation of energy.

T



$$\dot{F}_i = m \cdot \ddot{q} \quad (1687)$$

$$\frac{dP_i}{dt} = F_i \quad \text{--- ①}$$

$$L = K - U = \sum_i \frac{1}{2} m(\dot{q}_i)^2 - U(q_i) - \dot{q}_i(q_i) \dot{q} \quad (1760)$$

$$\frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) - \frac{\partial L}{\partial q_i} = 0 \quad \text{--- ②}$$

$$\frac{d}{dt} \left(m \dot{q}_i \right) + \left(+ \frac{\partial U}{\partial q_i} \right) = 0$$

$$\Rightarrow \ddot{q}_i = - \frac{1}{m} \nabla U(q)$$

$$\ddot{q}_i = - \frac{1}{m} \left(\frac{\partial U}{\partial q_i} \right) \sim ①$$

$$\begin{aligned} \frac{dL}{dt} &= \sum_i \underbrace{\frac{\partial L}{\partial q_i} \frac{d\dot{q}_i}{dt}}_{(1)} + \underbrace{\frac{\partial L}{\partial \dot{q}_i} \frac{d\dot{q}_i}{dt}}_{(2)} \\ &= \frac{dL}{d\dot{q}_i} \frac{d}{dt} (\dot{q}_i) \\ (1) &= \sum_i \frac{\partial L}{\partial \dot{q}_i} \frac{d\dot{q}_i}{dt} = \sum_i \left(\frac{\partial \dot{q}_i}{\partial q_i} \right) \frac{\partial L}{\partial q_i} \left(\frac{d\dot{q}_i}{dt} \right) \\ &= \sum_i \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) \dot{q}_i \\ \Rightarrow \frac{d}{dt} \left[\underbrace{\frac{\partial L}{\partial \dot{q}_i} \dot{q}_i}_{(1)} \right] &+ \underbrace{\frac{d}{dt} \left[\frac{\partial L}{\partial q_i} \right]}_{(2)} \dot{q}_i \quad \text{RHS} \end{aligned}$$

$$\Rightarrow \frac{d}{dt} \left[\frac{\partial L}{\partial \dot{q}_i} \dot{q}_i \right] + \frac{d}{dt} \left[\frac{\partial L}{\partial q_i} \right] \dot{q}_i = 0 \quad \text{LHS}$$

$$\begin{cases} \frac{\partial L}{\partial \dot{q}_i} = p_i & \frac{\partial L}{\partial q_i} = \left(\frac{\partial \dot{q}_i}{\partial q_i} \right) \frac{\partial L}{\partial \dot{q}_i} = \frac{d}{dt} p_i = \dot{p}_i \\ \frac{\partial L}{\partial q_i} = \dot{p}_i & \end{cases} \quad \text{EOM L}$$

$$H = -L + p \dot{q}$$

$$\begin{aligned} dH &= dp \dot{q} + d\dot{q} p - \cancel{\frac{\partial L}{\partial \dot{q}_i} d\dot{q}_i} - \cancel{\frac{\partial L}{\partial q_i} dq_i} = \cancel{\dot{q} dp - \dot{p} dq} \\ &\quad \cancel{p dq} \quad \cancel{\dot{p} dq} \quad \frac{\partial H}{\partial p} \quad \frac{\partial H}{\partial q} \end{aligned}$$

$$① L \leftrightarrow H$$

$$② \frac{dH}{dt} = 0$$

$$e^{-\beta t}$$

$$V \in C^\infty$$

$$\begin{aligned} EOM & \quad 1/ \\ \frac{\partial H}{\partial p} &= \dot{q} \\ \frac{\partial H}{\partial q} &= -\dot{p} \end{aligned}$$

$$S = k_B N \left[\log \left(\frac{V}{N} \left(\frac{4\pi m E}{3k_B T} \right)^{3/2} \right) + \frac{5}{2} \right]$$

: S.T. eq.
Sackur-Tetrode
 $S(N, V, E)$

$$(a) E = \frac{3k_B T}{4\pi m} \left(\frac{N}{V} \right)^{2/3} \exp \left[\frac{2S}{3k_B T} - \frac{5}{3} \right]$$

$$T = \left(\frac{\partial E}{\partial S} \right)_{N, V} \Rightarrow T = \frac{2}{3k_B T} E \Rightarrow E = \frac{3}{2} k_B T$$

$$P = \left(\frac{\partial E}{\partial V} \right)_{S, N} \Rightarrow P = \frac{2}{3} \frac{1}{V} E \Rightarrow PV = k_B T$$

$$\mu = \left(\frac{\partial E}{\partial N} \right)_{S, V} \Rightarrow \sim$$

$$(b) A = E - TS$$

$$= \frac{3}{2} N k_B T - \frac{1}{2} \left[\log \left(\left(\frac{2\pi m k_B T}{h^2} \right)^{3/2} \frac{N}{V} \right) + \frac{5}{2} \right] k_B T$$

(N, V, T)

$$(c) G = E - TS + PV$$

$$= -N k_B T \left[\log \left(\frac{k_B T}{P} \left(\frac{2\pi m k_B T}{h^2} \right)^{1/2} \right) \right] \quad (N, P, T)$$

$$= ?$$

$$(d) \frac{dG}{dT} = T \left(\frac{\partial S}{\partial T} \right)_{N, V} = \frac{3}{2} N k_B \quad (N, V, T)$$

$$C_V = \frac{dG}{dT} = T \left(\frac{\partial S}{\partial T} \right)_{N, P} = \frac{5}{2} N k_B \quad (N, P, T)$$

$$C_P = \frac{dG}{dP} = T \left(\frac{\partial S}{\partial P} \right)_{N, T} = \frac{3}{2} N k_B \quad (N, P, T)$$

$$C_P - C_V = N k_B$$

$$(e) \alpha = \frac{1}{V} \left(\frac{\partial V}{\partial T} \right)_{P, U} \cdot \frac{i}{V} \frac{\partial}{\partial T} \frac{N k_B T}{P} = \frac{1}{V} \frac{N k_B}{P} = \frac{1}{T}$$

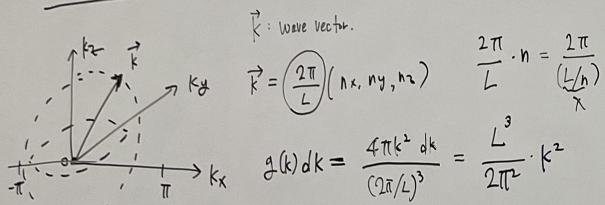
$$\beta = -\frac{1}{V} \left(\frac{\partial V}{\partial P} \right)_{U, T} = \frac{1}{V} \frac{N k_B T}{P^2} = \frac{1}{P}$$

$$Z = \sum_v e^{-\beta E(v)} \approx \int dE \Omega(E) e^{-\beta E(v)} \quad \text{Laplace}$$

$$= \int dE e^{-\beta \frac{E(v) - \log \Omega(E)}{A(v)}} \quad \text{Legendre}$$

Ω
 Z
 E
 A
 G
 Σ
 Δ
 T
 P

Density of States.

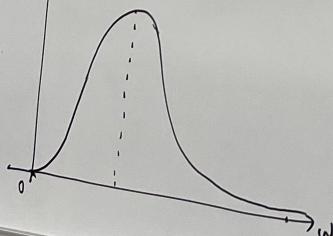


Planck's law

$$u \propto \langle \varepsilon(\omega) \cdot g(\omega) d\omega \rangle_B = \frac{1}{e^{\beta\hbar\omega} - 1} \cdot \omega \cdot \omega^2 d\omega$$

$$\propto \frac{\omega^3}{e^{\beta\hbar\omega} - 1}$$

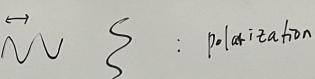
$$\begin{aligned} u &\propto \omega^3 \\ u &\rightarrow \text{Rayleigh-Jeans law.} \\ n &\rightarrow 0 \quad U \rightarrow 0 \end{aligned}$$



photons

$e \uparrow$

\vec{P}



$$p = \frac{h}{\lambda} = \frac{hf}{c} \Rightarrow pc = \epsilon$$

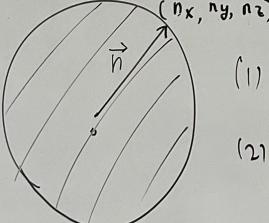
$$k = \frac{2\pi}{\lambda}$$

$$\epsilon = hf = \hbar\omega$$

$$p = \hbar k$$

$$\|P\| \propto \hbar \|\vec{n}\|$$

$$\Rightarrow \|\vec{n}\| \propto \|P\|$$



$$(1) N = \frac{4}{3}\pi \|\vec{R}\|^3 \propto \|P\|^3 \propto \|\epsilon\|^3 \propto \omega^3 \rightarrow g(\omega) d\omega \propto \omega^2 d\omega$$

$$(2) 4\pi \|\vec{R}\|^2 \cdot d\|\vec{n}\| \propto \omega^2 d\omega = g(\omega) d\omega$$

$$g(\omega) d\omega \propto \omega^2 d\omega$$

• Grand Canonical Ensemble.

$$Z_{\text{Boson}} = \sum_{n_0=0}^{\infty} \sum_{n_1=0}^{\infty} \dots \sum_{n_N=0}^{\infty} e^{-\beta(\sum_j n_j \epsilon_j - \mu \sum_j \epsilon_j)}$$

ϵ_N

\vdots

ϵ_j

ϵ_1

ϵ_0

$\uparrow \mu$

(μ) $V T$

$$= \prod_{j=0}^N \frac{1}{\prod_{i=0}^j e^{-\beta(\epsilon_i - \mu)}} \quad \left\{ \begin{array}{l} \sum_{n=0}^{\infty} r^n = \frac{1}{1-r} \\ |r| < 1 \end{array} \right.$$

$$\rightarrow \frac{1}{1 - e^{-\beta(\epsilon_0 - \mu)}}$$

$$\begin{cases} -\beta(\epsilon_j - \mu) < 0 \rightarrow \epsilon_j > \mu \\ \end{cases}$$

$$Z_B = \prod_{j=0}^N \frac{1}{1 - e^{-\beta(\mu + \epsilon_j)}}$$

$$Z_F = \prod_{j=0}^N 1 + e^{-\beta(\epsilon_j - \mu)}$$

$$\langle n_j \rangle_{\text{Bosons}} = \frac{\partial \log Z_B}{\partial (-\beta(\epsilon_j - \mu))}$$

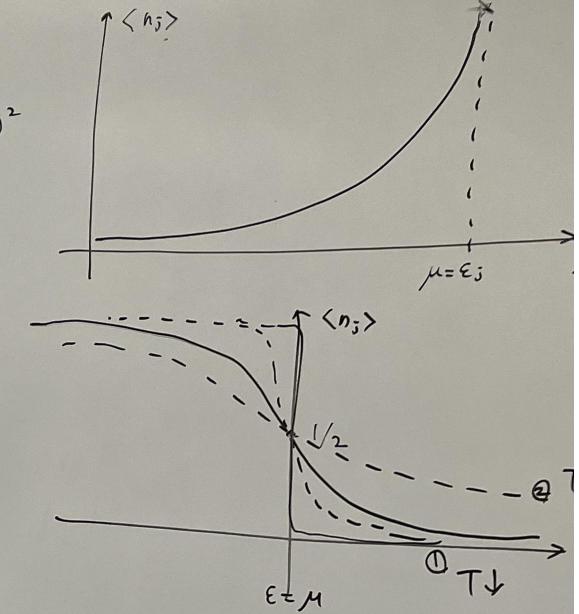
$$= \frac{1}{e^{\beta(\epsilon_j - \mu)} - 1}$$

$$\Delta h_j \text{ Bosons} = \frac{\partial^2 \log Z_B}{\partial (-\beta(\epsilon_j - \mu))^2}$$

$$\langle n_j \rangle_{\text{Fermion}} = \frac{1}{e^{\beta(\epsilon_j - \mu)} + 1}$$

$$\Delta h_j \text{ F.} \quad \beta = \frac{1}{k_B T}$$

$$\rightarrow \frac{\partial \log Z}{\partial (-\beta)} = \frac{1}{Z} \frac{\partial Z}{\partial (-\beta)}$$



critical phenomena

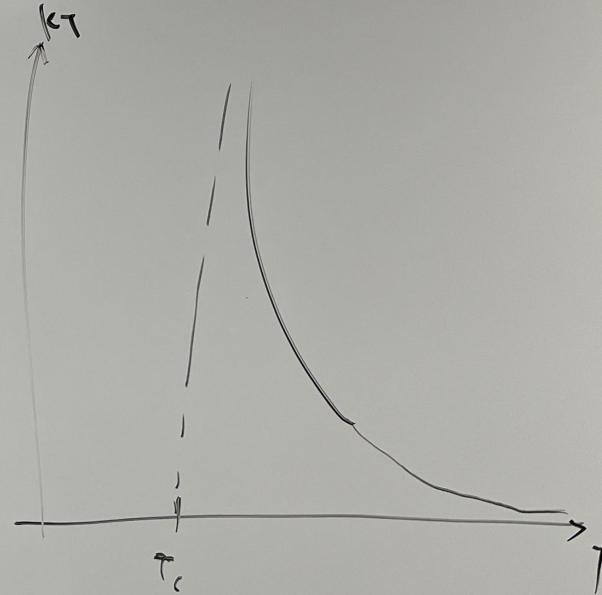
$$\begin{aligned}\hat{P} &= P/P_c \\ \hat{V} &= V/V_c \\ \hat{T} &= T/T_c\end{aligned} \quad \left. \right\} \rightarrow \left(\hat{P} + \frac{3}{\hat{V}^2} \right) \cdot \left(3\hat{V} - 1 \right) = 8\hat{T}$$

$$-\frac{\partial P}{\partial V} = -k_B T (V-b)^{-2} + 2\alpha V^{-3}$$

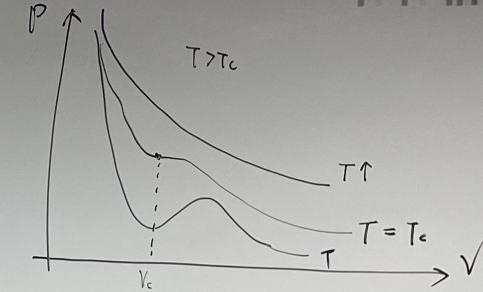
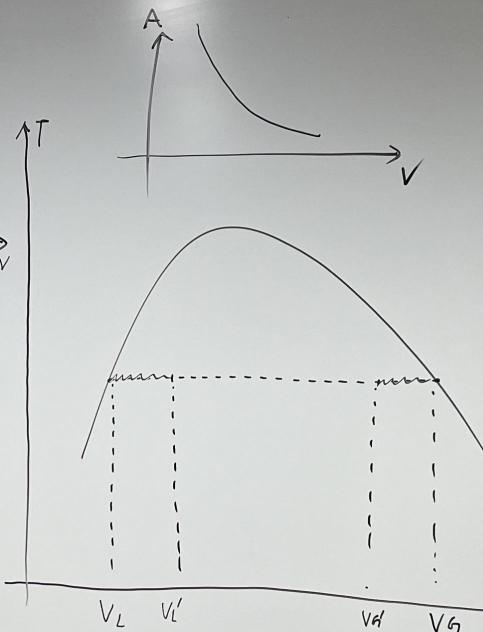
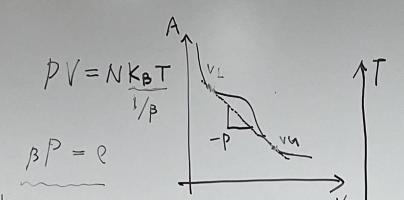
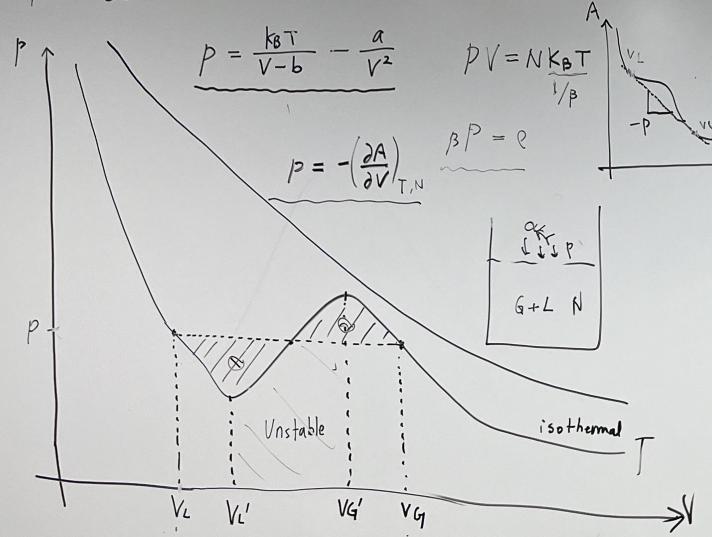
$$\frac{\partial V}{\partial P} = \frac{1}{-k_B T (V-b)^{-2} + 2\alpha V^{-3}}$$

$$k_T = -\frac{1}{V} \frac{\partial V}{\partial P} = \frac{1}{k_B T \cdot V (V-b)^{-2} - 2\alpha V^{-2}}$$

$$= \frac{4b}{3k_B} (T - T_c)^{-1} \quad T > T_c \quad \alpha = 1$$



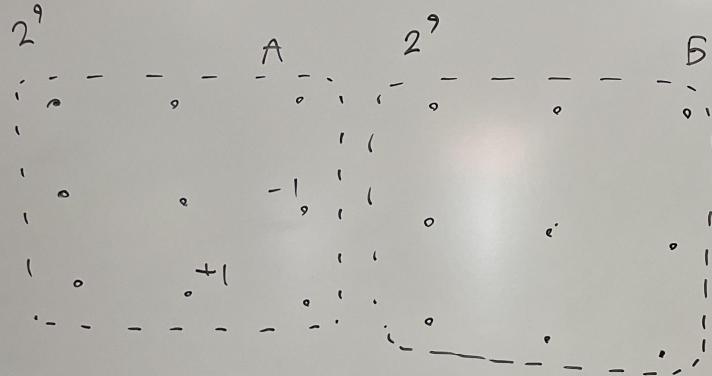
• Van der Waals Model



$$\textcircled{1} \quad \left. \frac{\partial P}{\partial V} \right|_{V_c} = 0 \Rightarrow 2\alpha V_c^{-3} = k_B T_c (V_c - b)^{-2}$$

$$\textcircled{2} \quad \left. \frac{\partial^2 P}{\partial V^2} \right|_{V_c} = 0 \Rightarrow 6\alpha V_c^{-4} = 2k_B T_c (V_c - b)^{-3}$$

$$\boxed{\begin{aligned} V_c &= 3b \\ k_B T_c &= \frac{8a}{27b} \\ P_c &= \frac{a}{27b^2} \end{aligned}}$$



$$18 \text{ spins} = 2^{18}$$

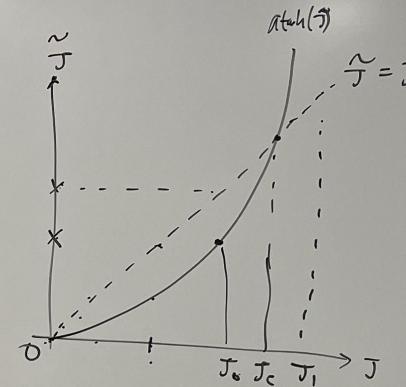
$$\sigma_A = \begin{cases} +1 & \sum_{i,j \in A} > 0 \\ -1 & " < 0 \end{cases}$$

$$\tilde{\mathcal{Z}} = \sum_{\sigma_A, \sigma_B} \tilde{z}(\sigma_A, \sigma_B)$$

$$\tilde{z} = e^{-\beta \tilde{H}} \quad \tilde{H} = -\tilde{J} \sigma_A \sigma_B - C - \tilde{h} \sigma_A - \tilde{h} \sigma_B$$

MATH.

$$\beta \tilde{J} = \operatorname{arctanh}(\beta J) \quad \begin{pmatrix} \tilde{z}_{+1,+1} & \tilde{z}_{+1,-1} \\ \tilde{z}_{-1,+1} & \tilde{z}_{-1,-1} \end{pmatrix} = \begin{pmatrix} e^{\beta \tilde{J} + \beta z h} & e^{-\beta \tilde{J}} \\ e^{\beta \tilde{J}} & e^{\beta \tilde{J} - \beta z h} \end{pmatrix}$$



$$\begin{aligned} & \frac{\tilde{z}_{+1,+1} - \tilde{z}_{+1,-1}}{\tilde{z}_{+1,+1} + \tilde{z}_{+1,-1}} \\ &= \frac{e^{\beta \tilde{J} + \beta z h} - e^{-\beta \tilde{J} - \beta z h}}{e^{\beta \tilde{J} + \beta z h} + e^{-\beta \tilde{J} - \beta z h}} \end{aligned}$$

$$= \tanh\left(\frac{\beta \tilde{J} + \beta z h}{\tilde{J}}\right) = \frac{\beta J}{\tilde{J}} + \beta h.$$

Monte Carlo

$$\left\{ \begin{array}{l} \textcircled{1} \quad v \\ \textcircled{2} \quad v' \downarrow \Delta E \\ \textcircled{3} \quad \frac{P(v')}{P(v)} = \frac{e^{-\beta E(v')}}{e^{\beta E(v)}} = e^{-\beta(E(v') - E(v))} = e^{-\beta \Delta E} \\ \textcircled{4} \quad \text{Accept/Reject} \\ P_{\text{acc}}(v \rightarrow v') = \min[1, e^{-\beta \Delta E}] = 0.1 \end{array} \right. \text{Metropolis-Hastings.}$$

Detailed Balance \equiv No net flow

\downarrow
Equilibrium \leftarrow Reversible

$$P(v) \cdot P(v \rightarrow v') = P(v') \cdot P(v' \rightarrow v)$$

\approx Flux

$$P(v \rightarrow v') = P_{\text{gen}}(v \rightarrow v') P_{\text{acc}}(v \rightarrow v')$$

||
 $P_{\text{gen}}(v' \rightarrow v)$ (if symmetric)

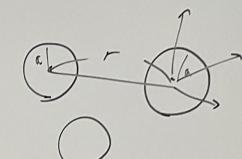
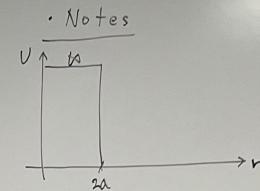
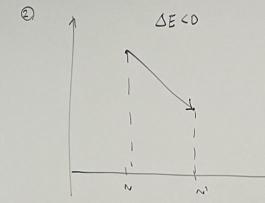
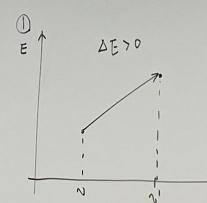
Python - How to prob.

~~import random~~
 $[0, 1] \xrightarrow{\text{if }} P < 1$
 if $\text{np.random.rand()} < P$

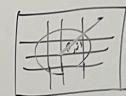


D.B. for M.C.

$$\frac{P(v')}{P(v)} = \frac{P(v \rightarrow v')}{P(v' \rightarrow v)} = \frac{\min(1, e^{-\beta \Delta E})}{\min(1, e^{+\beta \Delta E})}$$



$$|r_{ij} - r_{i,j+1}|^p$$



$$\frac{P(v')}{P(v)} = \frac{e^{-\beta \Delta E}}{1} = e^{-\beta \Delta E}$$

$$\frac{P(v')}{P(v)} = \frac{1}{e^{\beta \Delta E}} = e^{-\beta \Delta E}$$

PF3.

$$\vec{k} = (k_x, k_y, k_z) = \left(\frac{2\pi}{L}, (n_x, n_y, n_z) \right) \quad E_0 = mc^2$$

$$\underline{\Sigma} = \left(p^2 c^2 + \frac{m^2 c^4}{E_0^2} \right)^{1/2} \quad \vec{p} = h \vec{k} \quad p^2 c^2 = h^2 k^2 c^2 \quad p^2 = h^2 k^2$$

(a) $g(E)$

$$E^2 - E_0^2 = h^2 k^2 c^2 \quad (1)$$

(b) $E - E_0 \ll E_0$

$$2E dE = h^2 c^2 k dk$$

(c) $E - E_0 \gg E_0$

$$\frac{dE}{dk} = \frac{h^2 c^2}{E} k \rightarrow dk/dE = \frac{E}{h^2 c^2} \frac{1}{k}$$

$$g(E) dE = dN$$

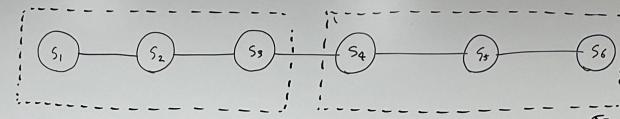
$$\underline{g(E)} = \frac{dN}{dE} = \frac{dN}{dk} \cdot \frac{dk}{dE}$$



$$\therefore g(E) \propto \frac{1}{E^2 - E_0^2}$$

$$\begin{aligned} dN &= 4\pi k^2 dk \cdot \left(\frac{L}{2\pi} \right)^3 \\ N &= \frac{4}{3}\pi k^3 \left(\frac{L}{2\pi} \right)^3 \\ N &= \frac{4}{3}\pi k^3 L^3 \end{aligned}$$

PF5.



$$H = -J \sum_{i,j>} S_i S_j \quad (h=0)$$

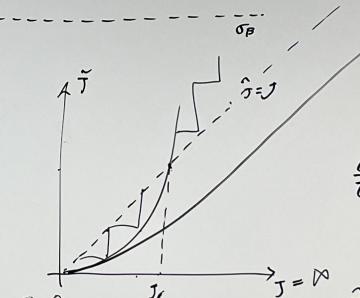
$$\sigma_A = \begin{cases} +1 & S_1 + S_2 + S_3 > 0 \\ -1 & S_1 + S_2 + S_3 < 0 \end{cases}$$

$$\sigma_B = \begin{cases} +1 & S_4 + S_5 + S_6 > 0 \\ -1 & S_4 + S_5 + S_6 < 0 \end{cases}$$

(Q) Find J_c

$$\tilde{H} = -\tilde{J} \sigma_A \sigma_B - C$$

$$\begin{array}{ccc} 1) & +1 & +1 \\ & +1 & -1 \\ & -1 & +1 \\ & -1 & -1 \end{array} \rightarrow 16 \quad \tilde{z} (\sigma_A = +1, \sigma_B = +1) \\ \begin{array}{ccc} & +1 & -1 \\ & -1 & +1 \\ & -1 & -1 \end{array} \rightarrow 16 \quad \tilde{z} (\sigma_A = +1, \sigma_B = -1) \\ \begin{array}{ccc} & +1 & +1 \\ & -1 & -1 \\ & -1 & -1 \end{array} \rightarrow 16 \quad \tilde{z} (\sigma_A = -1, \sigma_B = +1) \\ \begin{array}{ccc} & +1 & -1 \\ & -1 & -1 \\ & -1 & -1 \end{array} \rightarrow 16 \quad \tilde{z} (\sigma_A = -1, \sigma_B = -1) \end{array}$$



$$\tilde{z} (+1, +1) = \sum_{S>1} z(S_1 \sim S_6)$$

$$\tilde{z} (+1, -1) = \frac{P_J}{?}$$

$$\tilde{J} = f(J)$$

$$J_c$$

3 - (b) $E - E_0 \ll E_0$

$$f(E) \propto E \sqrt{E^2 - E_0^2} = E \sqrt{(E - E_0)(E + E_0)} \\ E = \sqrt{p^2 c^2 + m^2 c^4} = E_0 \sqrt{1 + \left(\frac{pc}{E_0}\right)^2} = E_0 \left(1 + \frac{1}{2} \left(\frac{pc}{E_0}\right)^2\right) = E_0 \left(1 + \frac{1}{2} \cdot \left(\frac{P}{mc}\right)^2\right)$$

$$E = E_0 \left(1 + \frac{1}{2} \left(\frac{P}{mc}\right)^2\right)$$