

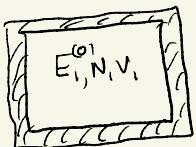
L7 Recap

- * Want to understand macro ($N \gg 1$) systems from microscopic DOF + dynamics (i.e., derive/explain Thermodynamics)
- * Impractical to solve Schrödinger/Hamilton's eqns. AND know all initial conditions for (say) 10^{23} particles
- * Many-to-1 relationship of micro- + macrostates
 - e.g., Huge # micro $(\vec{q}_1, \dots, \vec{q}_N, \vec{p}_1, \dots, \vec{p}_N)$ configs consistent with a given macro (E, V, N) .
- * Since our measurement/knowledge of the macrostate cannot distinguish between the huge # of microstates that are consistent with (E, V, N) , the simplest and most unbiased assumption is that all such microstates are equally probable
- * Multiplicity function $\mathcal{N}(E, V, N) \equiv \# \text{ of microstates that give } E, V, N$

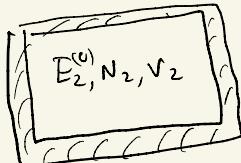
"Equal a priori probability postulate"

$$\left| \begin{array}{l} p(i) = \frac{1}{\mathcal{N}(E, V, N)} \quad \text{for } i = \text{one of the } \mathcal{N}-\text{microstates} \\ \qquad \qquad \qquad = 0 \quad \text{for } i \neq " " \end{array} \right.$$

Relation to Entropy



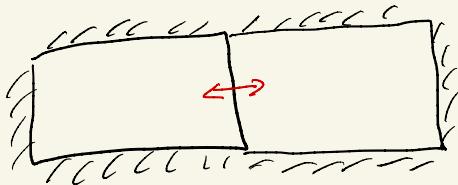
$$\mathcal{N}_1(E_1^{(o)}, N_1, V_1)$$



$$\mathcal{N}_2(E_2^{(o)}, N_2, V_2)$$



put in thermal contact
so can exchange heat w/ each other



$E_1 + E_2$ will change, but such that

$$E = E_1 + E_2 = E_1^{(o)} + E_2^{(o)} = \text{fixed}$$

$\mathcal{N}_{\text{TOT}}(E) =$ total # microstates for combined system
with E, N, V

$$= \sum_{E_1} \mathcal{N}_1(E_1) \mathcal{N}_2(E - E_1)$$

From experience we know if we wait long enough, $E_1 + E_2$ settle down to E_1^, E_2^*

\Rightarrow For large systems, find one particular term $E_1 = E_1^*$ dominates by a lot

$$\approx \mathcal{N}_1(E_1^*) \mathcal{N}_2(E - E_1^*)$$

E_1^* (+ hence $E_2^* = E - E_1^*$) determined by Maximizing

$$\frac{d}{dE_1} \left[\mathcal{N}_1(E_1) \mathcal{N}_2(E - E_1) \right]_{N, V} = \bigcirc$$

* algebra ... (see last class)

$$\Rightarrow \left. \frac{d}{dE_1} (\ln \Lambda_1) \right|_{N_1, V_1, E_1 = E^*} = \left. \frac{d}{dE_2} \ln \Lambda_2 \right|_{N_2, V_2, E_2 = E^*} = \beta$$

* So if 2 isolated systems are put into thermal contact so they can exchange energy (heat), they eventually settle down so that $\beta_1 = \beta_2 = \beta$

* This tells us β has something to do w/ temperature, and since TD tells us $\left. \frac{\delta S}{\delta E} \right|_{N, V} = \frac{1}{T}$, we expect $S + \ln \Lambda$ are equal up to a universal constant (which we know from experiment is k_B)

$$\Rightarrow S(E, N, V) = k_B \log \Lambda(E, N, V) \quad ***$$

Side note: I find it very bizarre & interesting that we related S , a quantity that was invented thinking about heat engines, to the # of microstates (& ultimately QM) States!

$$\Rightarrow \left[\beta = \frac{1}{k_B T}; \gamma = \frac{\mu}{T}, \xi = -\frac{\mu}{T} \right]$$

*Example: N Spin $\frac{1}{2}$ particles in external B-field (+ take $B \parallel \hat{z}$)

$$E = -B \sum_{j=1}^N \sigma_{z,j} \quad \begin{array}{l} (\text{absorbed } g=2 + M_B \text{ Bohr Magneton} \\ \text{into def. of } B) \end{array}$$

$$= -2BS_z^{\text{tot}} \quad S_z^{\text{tot}} = \sum_{i=1}^N S_z^{(i)}$$

$$\Rightarrow E = \{-NB, -(N-2)B, -(N-4)B, \dots, (N-2)B, NB\}$$

$N+1$ total energy values (macrostates)

2^N total microstates

Note: Macrostate can be equivalently labelled
by E or $S_z^{\text{tot}} = S$ value (Bad notation alert!
Don't confuse this w/
entropy)

Find $N(S)$: Note that $S = \frac{1}{2} \sum_{j=1}^N \sigma_{z,j}$

$$= \frac{N_\uparrow - N_\downarrow}{2}$$

$$= N_\uparrow - \frac{N}{2}$$

\therefore Finding $N(S) \Rightarrow$ Count the number
of ways of choosing
 N_\uparrow out of N to have spin \uparrow .

* imagine randomly picking particles & placing them in a row with 1st N_\uparrow in up-state & rest N_\downarrow in down state



* $N!$ ways to arrange particles in row

* But a state w/ the same N_\uparrow particles in the 1st bin is the same state no matter which order the particles are in

\therefore "Overcount" by $N_\uparrow! N_\downarrow!$

$$\Rightarrow \mathcal{N}(N_\uparrow, N_\downarrow) = \frac{N!}{N_\uparrow! N_\downarrow!} = \frac{N!}{N_\uparrow! (N-N_\uparrow)!}$$

$$\text{but recall } S = N_\uparrow - \frac{N}{2} \Rightarrow N_\uparrow = S + \frac{N}{2}$$

$$\therefore \mathcal{N}(S, N) = \frac{N!}{\left(\frac{N}{2} + S\right)! \left(\frac{N}{2} - S\right)!}$$

or, since $E = -2BS$

$$\mathcal{N}(E, N) = \frac{N!}{\left(\frac{N}{2} + \frac{E}{2B}\right)! \left(\frac{N}{2} - \frac{E}{2B}\right)!}$$

* Keeping in mind that we are interested in macroscopic systems w/ $N \sim 10^{23}$, it's useful to have an analytic approx for $N!$ @ large $N \gg 1$.

Stirling's approximation:

$$\ln N! \sim N \ln N - N + \frac{1}{2} \ln(2\pi N) + \frac{1}{12N} + O\left(\frac{1}{N^2}\right) \dots$$

~~~~~ ~~~~~ ~~~~~  
 Leading piece      Sub-leading often dropped      can always ignore for  $N \sim 10^{23}$

### Simple proof for leading piece

$$\begin{aligned}
 \ln N! &= \ln[N \cdot (N-1) \cdot (N-2) \dots 1] \\
 &= \ln N + \ln(N-1) + \ln(N-2) \dots + \ln 1 \\
 &= \sum_{n=1}^N \ln n \approx \int_1^N \ln n \, dn \\
 &\stackrel{\ln n + n}{=} (n \ln n - n) \Big|_1^N \\
 &= N \ln N - N + 1 \\
 &\approx N \ln N - N
 \end{aligned}$$

\* See Pathria & Beale appendix (or maybe HW #2?) for a more sophisticated derivation including subleading terms via Steepest Descents method.

$$\therefore \mathcal{V}(S, N) = \frac{N!}{\left(\frac{N}{2}+S\right)! \left(\frac{N}{2}-S\right)!}$$

$$\Rightarrow \ln \mathcal{V}(S, N) = \ln N! - \ln \left[ \left( \frac{N}{2} + S \right)! \right] - \ln \left[ \left( \frac{N}{2} - S \right)! \right]$$

$$\stackrel{N \gg 1}{\approx} N \ln N - N - \left( \frac{N}{2} + S \right) \ln \left( \frac{N}{2} + S \right) + \cancel{\left( \frac{N}{2} + S \right)}$$

$$- \left( \frac{N}{2} - S \right) \ln \left( \frac{N}{2} - S \right) + \cancel{\left( \frac{N}{2} - S \right)}$$

$$= N \ln N - \left( \frac{N}{2} + S \right) \ln \left[ \frac{N}{2} \left( 1 + \frac{2S}{N} \right) \right]$$

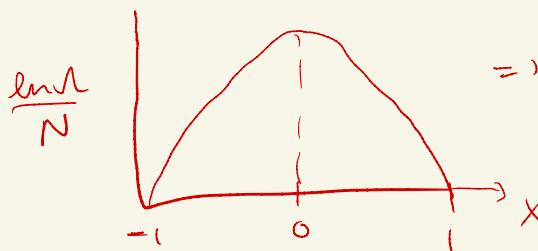
$$- \left( \frac{N}{2} - S \right) \ln \left[ \frac{N}{2} \left( 1 - \frac{2S}{N} \right) \right]$$

$$= N \ln N - \frac{N}{2} \left( 1 + \frac{2S}{N} \right) \ln \left[ \frac{N}{2} \left( 1 + \frac{2S}{N} \right) \right]$$

$$- \frac{N}{2} \left( 1 - \frac{2S}{N} \right) \ln \left[ \frac{N}{2} \left( 1 - \frac{2S}{N} \right) \right]$$

$$\text{let } x = \frac{2S}{N} \quad (x \in [-1, 1])$$

$$\boxed{\ln \mathcal{V} \approx -N \left[ \left( \frac{1+x}{2} \right) \ln \left( \frac{1+x}{2} \right) + \left( \frac{1-x}{2} \right) \ln \left( \frac{1-x}{2} \right) \right]}$$



$\Rightarrow \ln \mathcal{V}$  will be sharply peaked @  $x \approx 0$

\* Note:  $\ln \mathcal{N}(x=0) = N \ln 2 \Rightarrow \mathcal{N}(x=0) = 2^N$

which is a bit strange since  $2^N$  is the total # of microstates possible.

\* This is an artifact of the leading Stirling approx., but it also makes the point that the probabilities are dominated by a single term when  $N \rightarrow \infty$ .

\* If we kept the subleading term

$$\ln N! \approx N \ln N - N + \frac{1}{2} \ln(2\pi N) \quad ,$$

would find  $\mathcal{N}(x=0) \approx \left(\frac{2}{\pi N}\right)^{1/2} 2^N$

### Sharpness of $\mathcal{N}$

\* Even though probabilistic arguments underly stat. mech., we know for Macroscopic ( $N \sim 10^{23}$ ) systems, SM makes essentially exact statements (i.e., fluctuations of macro state variables are negligible)

\* We'll see this is a consequence of how sharply peaked  $\mathcal{N}$  becomes for large  $N$ .

\* Start from:

$$\ln \mathcal{N} \approx -N \left[ \left( \frac{1+x}{2} \right) \ln \left( \frac{1+x}{2} \right) + \left( \frac{1-x}{2} \right) \ln \left( \frac{1-x}{2} \right) \right]$$

\* analyze this in the limit  $N \gg 1$  and  $|x| = \left| \frac{2S}{N} \right| \ll 1$

$$\Rightarrow \ln \left( \frac{1+x}{2} \right) \approx -\ln 2 + x - \frac{x^2}{2} + \mathcal{O}(x^3)$$

$$\ln \left( \frac{1-x}{2} \right) \approx -\ln 2 - x - \frac{x^2}{2} + \mathcal{O}(x^3)$$

$$\Rightarrow \ln \mathcal{N} \approx -N \left[ \left( \frac{1+x}{2} \right) \left( -\ln 2 + x + \dots \right) + \left( \frac{1-x}{2} \right) \left( -\ln 2 - x + \dots \right) \right]$$

$$= -N \left[ -\ln 2 + \frac{x^2}{2} + \mathcal{O}(x^4) \right] = \ln 2^N - \frac{Nx^2}{2}$$

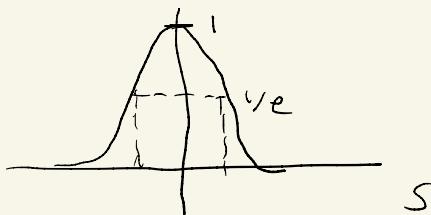
$$\Rightarrow \boxed{\mathcal{N}(N, x) \approx e^{\ln 2^N} x e^{-\frac{Nx^2}{2}} \\ = \mathcal{N}(N, 0) e^{-\frac{Nx^2}{2}}}$$

or in terms of  $S = \frac{Nx}{2}$

$$\Rightarrow \boxed{\mathcal{N}(N, S) = \mathcal{N}(N, 0) e^{-2S^2/N}}$$

∴ Fractional width (i.e., when  $\mathcal{N}(N,S)$  drops by  $1/e$   
compared to  $\mathcal{N}(N,0)$ )

$$\frac{S}{N} = \sqrt{\frac{1}{2N}}$$



e.g.:  $N = 10^{22} \Rightarrow \cancel{\sqrt{\frac{1}{2N}}} \sim 10^{-11}$

∴ The peak is very sharp  
for large  $N$

i.e., if  $P(S) = \frac{\mathcal{N}(N,S)}{\mathcal{N}_{\text{TOT}}} = \frac{\mathcal{N}(N,S)}{2^N}$

then this is essentially a  $\delta$ -function  
as  $N \rightarrow \infty$ , ~~as~~

## Ensemble Theory

ensemble = imaginary set of copies of our original system  
 that can be in any macroscopically allowed microstate

assumed very large

- \* Task of SM is to find probability  $P_i$  that a randomly drawn copy from the ensemble is in microstate  $i$
- \* Knowing this, we can then calculate average quantities

$$\langle \mathcal{O} \rangle = \sum_i P_i \mathcal{O}(i) \quad \mathcal{O}(i) = \text{value of } \mathcal{O} \text{ in microstate } i$$

\* Later, when we discuss Ergodic systems, we'll see

$$\langle \mathcal{O} \rangle = \sum_i P_i \mathcal{O}(i) = \bar{\mathcal{O}} = \frac{1}{T} \int_0^T dt \mathcal{O}(t)$$

### Microcanonical Ensemble (every copy has fixed $(E, V, N)$ )

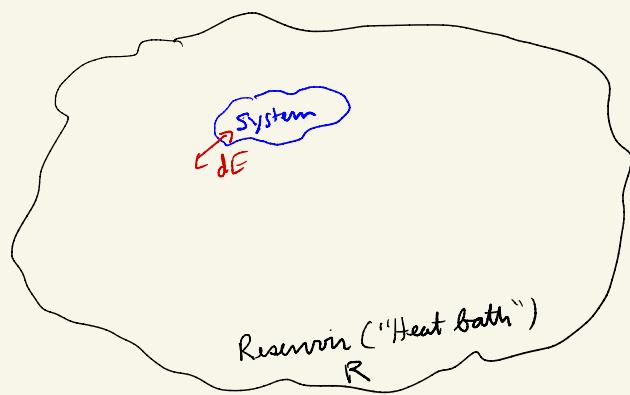
$$P_i = \frac{1}{N(E, V, N)} \quad \text{for } i \text{ giving fixed } (E, V, N)$$

$$\langle \mathcal{O} \rangle = \sum_i P_i \mathcal{O}(i) = \frac{1}{N} \sum_i \mathcal{O}(i)$$



Sum only over microstates  
 w/ fixed  $(E, V, N)$

# Canonical Ensemble (members at fixed T,V,N but E can fluctuate) [81]



- \* R is much bigger than System, though the latter is still microscopic
- \* Since R is so huge, can view it as holding S at fixed T.

$$E_s + E_R = E = \text{fixed} \quad (\text{together } S+R = \text{closed system})$$

\* What is  $p(i) = \text{prob. of system to be in } i^{\text{th}} \text{ microstate}$  ?



\* To answer this, treat  $(S+R)$  in microcanonical ensemble & consider 2 cases:

① S is in the particular microstate i

$$\mathcal{N}_R(E_R) \mathcal{N}_s(E_s=E_i) = \mathcal{N}_R(E-E_i) \cancel{\mathcal{N}_s(E_i)}^1 = \begin{matrix} \# \text{ of ways sys} \\ \text{can be in microstate} \\ i \end{matrix}$$

② S is in the particular microstate j

$$\mathcal{N}_R(E_R) \mathcal{N}_s(E_s=E_j) = \mathcal{N}_R(E-E_j) \cancel{\mathcal{N}_s(E_j)}^1 = \begin{matrix} \# \text{ of ways Sys} \\ \text{can be in} \\ \text{microstate } j \end{matrix}$$

\* Since the combined S+R is closed & treated in the microcanonical ensemble, all microstates equally likely

$$\therefore \frac{p(i)}{p(j)} = \frac{\frac{\# \text{ of ways sys}}{\text{can be in } i}}{\frac{\# \text{ of ways sys}}{\text{can be in } j}} = \frac{\mathcal{N}_R(E-E_i)}{\mathcal{N}_R(E-E_j)}$$

\* Since  $|E_i| \ll E$  by assumption that  $R \gg S$

$$\begin{aligned} \log[\mathcal{N}_R(E-E_i)] &\approx \log \mathcal{N}_R(E) - E_i \frac{\partial}{\partial E} \log \mathcal{N}_R + \dots \\ &= \log \mathcal{N}_R(E) - E_i \beta \end{aligned}$$

~~~~~  
ignore
(why?)
think
physically!

$$\Rightarrow \mathcal{N}_R(E-E_i) \approx e^{\log \mathcal{N}_R(E) - E_i \beta}$$

$$= \mathcal{N}_R(E) e^{-\beta E_i}$$

$$\Rightarrow \boxed{\frac{p(i)}{p(j)} = \frac{e^{-\beta E_i}}{e^{-\beta E_j}}}$$

$$\therefore p(i) = C e^{-\beta E_i} \quad \text{"Boltzmann Weight"}$$

Requiring $\sum_i p(i) = 1 \Rightarrow C = \frac{1}{\sum_j e^{-\beta E_j}}$

$$\Rightarrow p(i) = \frac{e^{-\beta E_i}}{\sum_j e^{-\beta E_j}} ; \quad \sum_j e^{-\beta E_j} \equiv Z$$

"partition function"

Note: We'll re-derive $p(i)$ & Z from a more ensemble based argument in a couple of lectures

Ensemble averages (canonical)

$$\left. \begin{aligned} \langle O \rangle &= \sum_i p(i) O(i) \\ &= \frac{1}{Z} \sum_j O_j e^{-\beta E_j} \end{aligned} \right\}$$

$$\text{ex: } \langle E \rangle = \frac{1}{Z} \sum_j E_j e^{-\beta E_j}$$

$$\begin{aligned} &= \frac{1}{Z} \frac{\partial}{\partial \beta} \sum_j e^{-\beta E_j} = -\frac{1}{Z} \frac{\partial}{\partial \beta} Z = -\frac{2}{\partial \beta} \log Z \\ &= T^2 \frac{\partial}{\partial T} \log Z \end{aligned}$$

*Note: $\langle E^2 \rangle - \langle E \rangle^2 \neq 0$

(fluctuations), but we'll show shortly that they vanish in $T \rightarrow 0$ limit.

* Connecting further to Thermodynamics, recall that
 Canonical ensemble \Rightarrow systems w/ fixed V, T, N

* Suggests looking at Helmholtz F

$$F = E - TS$$

$$dF = -SdT - PdV + \mu dN$$

$$\left(\frac{\partial F}{\partial T}\right)_{V,N} = -S; \quad \left(\frac{\partial F}{\partial V}\right)_{T,N} = -P; \quad \left(\frac{\partial F}{\partial N}\right)_{T,V} = \mu.$$

* From prior example,

$$\begin{aligned} \langle E \rangle &= -\frac{\partial \log Z}{\partial \beta} \Big|_{N,V} = F + TS \\ &\quad = F - T \left(\frac{\partial F}{\partial T} \right)_{V,N} \\ &\quad = F + \left(\frac{\partial F}{\partial \log \beta} \right)_{N,V} \\ &\quad = F + \left(\frac{\partial (\beta F)}{\partial \beta} \right)_{N,V} \end{aligned}$$

$$\therefore F = -\frac{\partial \log Z}{\partial \beta} - \frac{\partial (\beta F)}{\partial \beta}_{N,V}$$

** $\boxed{\Rightarrow F = -\frac{1}{\beta} \log Z}$ **