

We will take advantage of two properties
to instead develop a statistical
connection between microscopic & bulk
properties

Ensemble method & ergodic hypothesis

Definitions (reminder)

Macrostate - thermodynamic state of
the system eg N, V, E or N, V, T or n, p, T

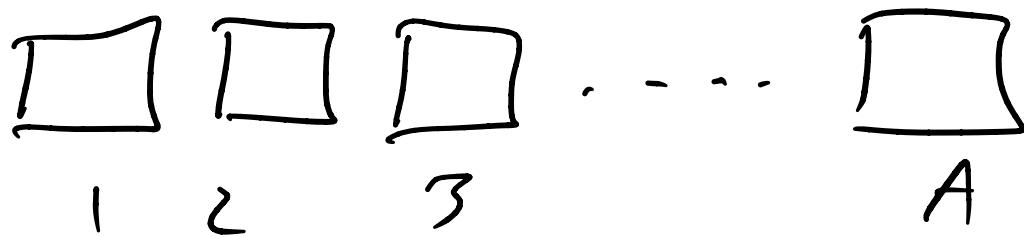
Microstate - $x_1, y_1, z_1, p_x, p_y, p_z$ of every atom
Configuration of system, c_g

Ergodicity - long enough time,
system explores all microstates

Spends "correct" amount of time in each
one (more later)

Ensemble method - if had many copies]
of system, average over copies gives } ergodic
Some result as $t \rightarrow \infty$ of one copy } hypothesis

Suppose a system has m possible states
A total copies of system



let N_i be the number in state i

$$\sum_{i=1}^m N_i = A$$

$$\{N_i\} = \{N_1, N_2, \dots, N_m\}$$

prob of each state $1-m$

$$\{p_i\} = \{p_1, \dots, p_m\} = \left\{ \frac{N_1}{A}, \frac{N_2}{A}, \dots, \frac{N_m}{A} \right\}$$

Ensemble average, of property y

$$\langle y \rangle_{\text{ens}} = \sum_{i=1}^m y_i p_i$$



Should be same as time avg $\langle y \rangle_t = \frac{1}{N_t} \sum_{j=1}^{N_t} y(t_j)$

An ensemble is a collection of copies
of a system all in the same macrostate

A "microcanonical" ensemble is one with
const N, V, E - isolated system obeying
Newton's equations $F=ma$, conserve E

How often does each microstate occur?
equilibrium maximizes entropy!

$S = k_B \ln W$, but what is $W(N, V, E)$
for an ensemble

N_1 copies in state 1, N_2 copies in state 2, ...
 $\sum_{i=1}^A N_i = A$

$$W = \frac{A!}{N_1! N_2! N_3! \dots N_m!} \quad (\text{multinomial})$$

for $A \gg N_i$ large

$$S = k \ln W = A \ln A - A - \sum_{i=1}^m (N_i \ln N_i - N_i)$$

adds to A
↓

$$S_0 \stackrel{def}{=} \sum_{i=1}^m N_i \ln N_i - \sum_{i=1}^m N_i$$

$$\frac{1}{k_B} \frac{\partial S}{\partial N_j} = 0 \quad \text{for all } N_j$$

$$\frac{1}{k_B} \frac{\partial S}{\partial N_j} = \frac{\partial(A \ln A)}{\partial N_j} - \frac{\partial}{\partial N_j} \left(\sum_{i=1}^m N_i \ln N_i \right)$$

$$A = \sum_{i=1}^m N_i$$

$$= A \frac{\partial \ln A}{\partial N_j} + \frac{\partial A}{\partial N_j} \ln A - \frac{N_j}{N_j} - \ln N_j$$

↑ only j matters

$$= 1 + \ln A - 1 - \ln N_j$$

$$= \ln A - \ln N_j$$

Smallest for N_j all maximized

But can't make all N_j max simultaneously

because $\sum N_j = m$, need
a constraint

Turns out (see pg 312-313)

Lagrange Multipliers for constraint

Instead, find where

$I = f(x) - \alpha (\text{constraint})$ is max

$$I = k \ln w - \alpha \left(\sum_{i=1}^m N_i - A \right) = 0$$

$$\frac{\partial I}{\partial N_j} = \ln A - \ln N_j - \alpha = 0 \quad \forall N_j$$

$$\alpha = \ln(A/N_j) \quad \text{is a constant}$$

$$\frac{N_j}{A} = e^{-\alpha} \quad \forall N_j$$

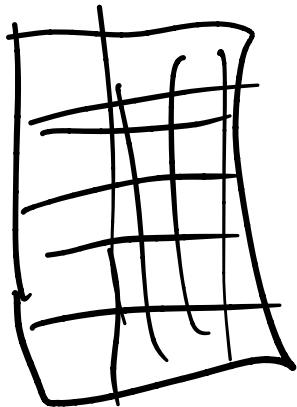
find λ by enforcing constraint

$$\sum_{i=1}^m N_i = A = \sum_{i=1}^m A e^{-\alpha} \Rightarrow e^\alpha = m$$

$$P_i = \frac{N_i}{\sum N_i} = \frac{1}{m}, \text{ equal distributions!}$$

Chapter 10 - Canonical Ensemble

Consider bunch of copies of system



N, V, E overall

know heat transfers until
here same temp

$$S = k \ln W$$

but Now

so for one system,

$$\tilde{N}, \tilde{V}, T$$

$$\text{where } N = A \tilde{N} \text{ a.s.m}$$

$$V = A \tilde{V}$$

$$E = \sum_{i=1}^m E_i N_i$$

Max S w/ E constraint & $\sum N_i = A$ sum
over states

$$S_{k_B} = \ln A - \alpha(\sum N_i - A) - \beta(\sum N_i \epsilon_i - \epsilon_T)$$

$$\frac{\partial S_{k_B}}{\partial N_j} = \ln \frac{A}{N_j} - \alpha - \beta \epsilon_j = 0$$

$$\text{and so } N_j = e^{-\alpha} e^{-\beta \epsilon_j} \cdot A$$

$$\sum_{j=1}^m N_j = A = A \sum_{j=1}^m e^{-\alpha} e^{-\beta \epsilon_j}$$

$$\text{so } e^\alpha = \sum_{j=1}^m e^{-\beta \epsilon_j}$$

this means $P_j = \frac{N_j}{A} = \frac{e^{-\beta \epsilon_j}}{\sum_{j=1}^m e^{-\beta \epsilon_j}}$

$$Q = \sum_{j=1}^m e^{-\beta \epsilon_j} \text{ is "partition function"}$$

for canonical ensemble $\delta e^{-\beta \epsilon_j}$

is the Boltzmann factor (relative weight)

formicano canonical ensemble

$$\beta F = 1 \quad (\text{like } \varepsilon_j = 0 \text{ or } \beta_j = 0)$$

$$\text{and } Q = \sum \beta F_i = \sum y_i = n$$

Average of y

$$\langle y \rangle = \sum_{i=1}^m y_i p_i = \sum_{i=1}^m y_i e^{-\beta \varepsilon_i} / \alpha$$

$$\varepsilon_g \langle \varepsilon \rangle = \sum_{i=1}^m \varepsilon_i e^{-\beta \varepsilon_i} / \alpha$$

Turns out I can show b) connection

to classical thermo that

$$\beta = 1/k_B T$$

so $Q = \sum_{i=1}^m e^{-\varepsilon_i / k_B T}$ & can get other thermo quantities from this function

$$\frac{\partial \ln X}{\partial \beta} = -\frac{1}{X} \frac{\partial X}{\partial \beta} \quad \text{if we consider}$$

$$-\frac{\partial \ln Q}{\partial \beta} = -\frac{1}{Q} \frac{\partial Q}{\partial \beta} = -\frac{1}{Q} \sum_{i=1}^m \varepsilon_i e^{-\beta \varepsilon_i}$$

$$U = - \frac{\partial \ln Q}{\partial \beta}$$

$$\text{also } \frac{\partial f}{\partial \beta} = \frac{\partial f}{\partial T} \frac{\partial T}{\partial \beta} = \frac{\partial f}{\partial T} \left(\frac{\partial \beta}{\partial T} \right)^{-1}$$

$$= - \left(\frac{1}{k_B T^2} \right) \frac{\partial f}{\partial T}$$

$$\text{so } U = k_B T^2 \frac{\partial \ln Q}{\partial T}$$

Gibbs Entropy

$$S = -k_B \sum p_i \ln p_i$$

$$= -k_B \sum p_i [-\beta E_i - \ln Q]$$

$$= k_B \beta \sum p_i E_i + k_B \ln Q$$

$$= k_B F + k_B \ln Q$$

$$T = \left(\frac{\partial u}{\partial S} \right)_V = \frac{\left(\frac{\partial u}{\partial \beta} \right)_V}{\left(\frac{\partial S}{\partial \beta} \right)_V}$$

$$\frac{\partial S}{\partial \beta} = k_B \left(1 + \beta \frac{\partial u}{\partial \beta} + \underbrace{\left(\frac{\partial \ln Q}{\partial \beta} \right)}_{= \zeta} \right)$$

$$= k_B \beta \left(\frac{\partial u}{\partial \beta} \right)_V$$

$$\therefore T = \frac{1}{k_B \beta} \Rightarrow \beta = \frac{1}{k_B T} \neq 0$$