

NVE:

$$S = k_B \ln \Omega(E)$$

NVT:

$$A = -k_B T \log Z$$

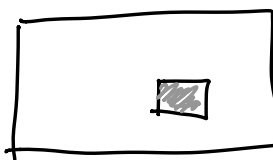
thermodynamics
e.g. macro-averages

Stat Mech.
microscopic degrees
of freedom

$E = \text{const}$



isolated
 $E(S, V, N)$



$T = \text{const}$

Leg. transform
 \Rightarrow

$$E - TS = A$$

$$-A = \max_S \{ TS - E \}$$

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

$$\Omega(E) = \int \frac{dx^N dp^N}{h^{3N}} = e^{S/k_B}$$

$$E - \Delta E \leq H(x, p) \leq E + \Delta E$$

Laplace

$$Z = \sum_i e^{-\beta E_i} = e^{-\beta A}$$

$$P(E) = \frac{1}{\Omega(E)}$$

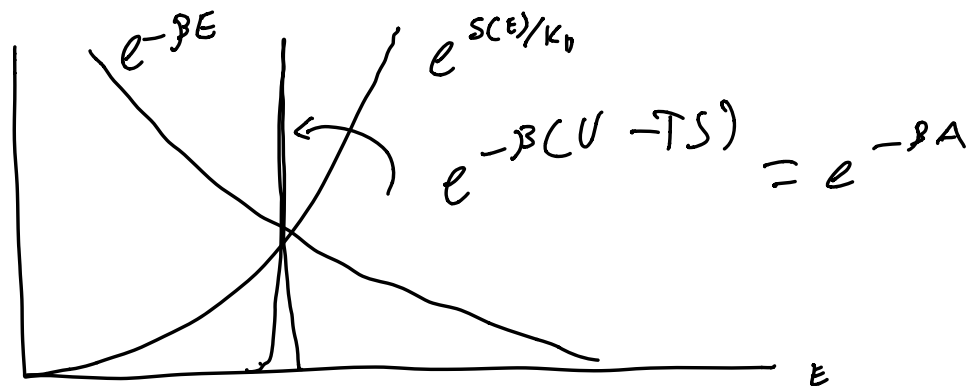
$$Z = \int \Omega(E) e^{-\beta E} dE$$

$$p = \frac{e^{-\beta E_i}}{Z}$$

$N \rightarrow \infty$

$$\int e^{-\beta(E - TS(E))} dE \approx e^{\beta(TS(E) - E)}$$

Large dev function $E \leftrightarrow A$



Thermodynamic equivalence of ensembles!!!

Gibbs formula of entropy

$$S = K_B \log \Omega = -K_B \log \frac{1}{\Omega}$$

$$TS = E - A$$

$$TS' = \sum p_i E_i + \frac{1}{\beta} \log Z$$

$$TS = -\frac{1}{\beta} \sum p_i \log e^{-\beta E_i} + \frac{1}{\beta} \log Z \cdot \sum_i p_i$$

$$TS = -\frac{1}{\beta} \sum p_i \log \frac{e^{-\beta E_i}}{Z}$$

$$S' = -K_B \sum p_i \log p_i \quad \text{Gibbs}$$

Shannon measure of information

SMI

n choice for a letter $\eta(n)$ increase

$$\eta(n) = c \log n$$

$$c = (\log 2)^{-1}$$

$$\eta(n) = \log_2 n$$

1 bit \equiv amount of info from single yes-no question

$$\eta(n) = \log_2 \frac{1}{p_n} = \log_2 \frac{1}{p_n}$$

$$\eta(n) = - \sum_{i=1}^n p_i \log p_i$$

example horses, roommate-keys, coins, die, etc.

$\log_2 27 \approx 4.755$ but real is 1.3 bits

$\frac{70}{250} \cdot \log_2 27 \approx 1.3$ bits

70 given to guess
250 long paragraph

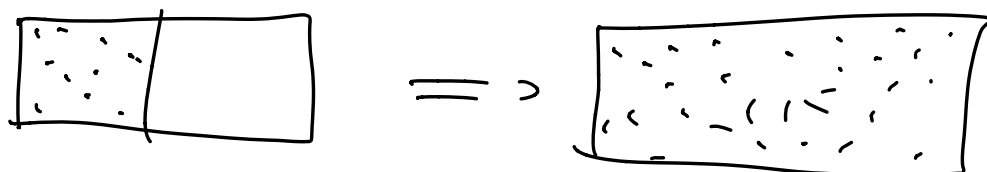
two kittens: arrival of new info $\log_2 4 - \log_2 3$
 Monty python example: $-\frac{1}{5} \log \frac{1}{5} - \frac{4}{5} \log \frac{4}{5}$
 ~ 66 bits

Units

$$1 \text{ bit/molecule} = k_B \log 2 = 9.57 \cdot 10^{-24} \text{ J/K.mol}$$

$$1 \text{ bit/mole} = R \log 2 = 5.7628 \text{ J} \cdot \text{K}^{-1} \text{ mol}^{-1}$$

$$1 \text{ J} \cdot \text{K}^{-1} \text{ mol} = 0.17352 \text{ bit/molecule}$$

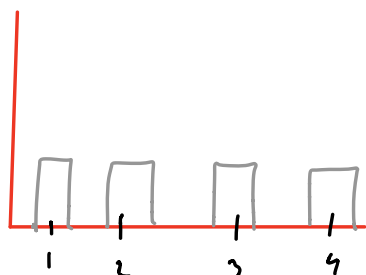


$$\Delta S = R \log 2 \quad \text{entropy increase?} \quad N_A \text{ bits}$$

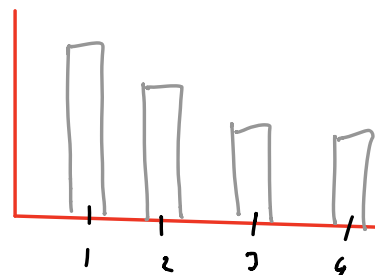
$$S = - \sum p_i \log p_i$$

uncertainty over picking microstate
 or # Yes/No questions to identify state.

$$(1) S'(\text{flat dist}) > S'(\text{non-flat dist})$$



$$p_1 = p_2 = p_3 = p_4 = 1/4$$



$$p_1 > p_2 > p_3 > p_4$$

Why entropy increase?

$$f = f^* = 1/2$$

$$P(f) = e^{-N I(f)} \quad I(f) = -f \log f - (1-f) \log(1-f)$$

Large dev function is Entropy! or information!!

System tends to the greatest uncertainty \equiv highest prob

Entropy is a functional of probab distribution. Systems evolve towards prob dis over microstates that maximize their entropy given any constraints

Max Ent

Principle of insufficient reason
maximum entropy principle
fair allocation principle

$$\sum_i p_i E_i = E \quad \sum_i N_i p_i = N \quad \sum_i p_i = 1$$

$$S = - \sum_i p_i \log p_i - \lambda (\sum_i p_i - 1)$$

$$\frac{\partial S}{\partial p_i} = 0 \quad -1 - \log p_i - \lambda = 0$$

$$\log p_i = -1 - \lambda$$

$$p_i = e^{\lambda'} \quad \sum_i p_i = 1$$

$$\sum_i (e^{\lambda'}) = e^{\lambda'} \Omega$$

$$p_i = \frac{1}{\Omega} \quad \swarrow \text{ \# states }$$

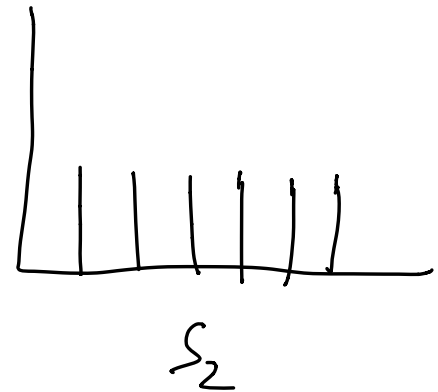
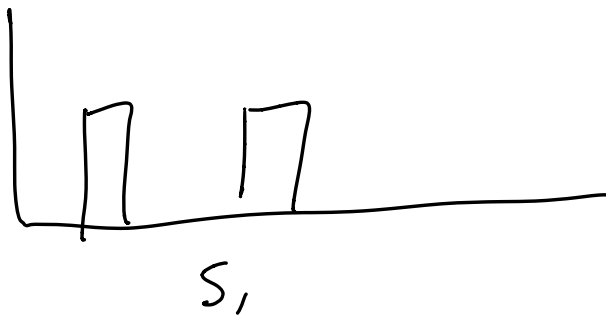
$$\sum_i p_i E_i = E \quad \sum_i p_i = 1$$

$$-1 - \log p_i - \lambda - \beta E_i$$

$$p_i \sim e^{-\beta E_i}$$

Boltzmann
distribution

constrained maximization



rigged coin and
 $\langle x \rangle = 0.8$

rigged die
 $\langle x \rangle \approx 5.5$

