

Jaynes (1957) : stat. phys. as an inference problem using limited data (such as  $\bar{E}$ )

Maximize  $S = -k_B \sum_i p_i \log p_i$

( $p_i$  - prob. to be in cell  $i$  of phase space)

with constraints:

$$(1) \sum_i p_i = 1 \text{ normal'n}$$

$$(2) \langle E \rangle \equiv \sum_i p_i E_i = \bar{E}$$

↑  
 energy of a  
System of particles

$$\delta \left[ \frac{S}{k_B} - \beta \left( \sum_i p_i E_i - \bar{E} \right) - \alpha \left( \sum_i p_i - 1 \right) \right] = 0$$

↑  
 wrt  $\{p_i\}$  &  $\alpha, \beta$

$$p_i^* = \frac{e^{-\beta E_i}}{\sum_j e^{-\beta E_j}} \} \equiv Q, \text{ part'n f'n}$$

$$S_{\max} = -k_B \sum_i p_i^* \log p_i^* = -k_B \left[ \sum_i p_i^* (-\beta E_i) - \underbrace{- \sum_i p_i^* (\sum_j e^{-\beta E_j})}_{\log Q} \right] =$$

$$= +k_B \log Q + \underbrace{k_B \beta \langle E \rangle}_{1/T} = \frac{\langle E \rangle - F}{T},$$

where

$F = -k_B T \log Q$  is the free en.

①

Shore & Johnson:

(1980, 1981)

Max of entropy  $\rightarrow$  self-consistent way to draw inferences from prob. distrib's.

Max of  $H$  with constraints should be:

- 1) unique
- 2) inv wrt coord. transform'n
- 3) relative prob. of two subsets of outcomes should not depend on other subsets if data are provided indep. ~~for~~ each subset
- 4) Prob. for 2 indep. systems should be a product of marginal probs.

Consider prob. prior probs.

$$H(\{p_i, q_i\}) - \lambda \left( \sum_i d_i p_i - \bar{d} \right)$$

$d$  - some property of interest

$p_i \rightarrow q_i$  when there're no priors constraints (i.e., no data)

Constraints / data lead to  $p_i \neq q_i$   $\downarrow$  discard  $\lambda \bar{d}$

Start with  $(\partial_{p_i} - \partial_{p_k}) [H - \lambda \sum_i d_i p_i]$

E.g.  $p_j$  increased,  $p_k$  decreased  $\Rightarrow$

$\Rightarrow$  other cells  $l \neq j, k$  unaffected

(axiom 3):

②

$$p_e (\partial p_j - \partial p_k) [H - \sum_i d_i p_i] = 0$$

$$\downarrow \quad l \neq k, j$$

$\downarrow$   $(\partial p_j - \partial p_k) H$  must depend only  
on  $j, k$

$\downarrow$   $\partial p_m H$  ~~must~~ must depend only  
on  $m$

$$H = \sum_i f(p_i, q_i)$$

"cell-level" quantities

Continuous repres'n:

$$H = \int Dx \ f(p(x), q(x))$$

Coord. inv. of the max (axiom 2):  
 $x \rightarrow y \Rightarrow Dy = Dx \cdot J$   $\uparrow$  Jacobian

$$\int Dx p(x) = 1 \Rightarrow \underbrace{\int Dy}_{Dx \cdot J} \underbrace{p(y)}_{\frac{p(x)}{J}} = 1$$

$$\int Dx p(x) a(x) = \int \underbrace{Dy}_{Dx \cdot J} \underbrace{p(y)}_{\frac{p(x)}{J}} a(y) = \bar{a}$$

$\Downarrow$

$$a(x) = a(y)$$

$$\text{So, } H' = \int Dx J f\left(\frac{p(x)}{J}, \frac{q(x)}{J}\right)$$

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$$\frac{\delta}{\delta p(x)} \left[ H - \lambda \int \mathcal{D}x \, p(x) d(x) \right] = 0 \quad \text{gives}$$

$$\underbrace{\frac{\partial f[p(x), q_f(x)]}{\partial p(x)}}_{\text{" } g[p(x), q_f(x)]} - \lambda d(x) = 0.$$

$$\frac{\delta}{\delta p(y)} \left[ H' - \lambda' \int \mathcal{D}y \, p(y) d(y) \right] =$$

$$= g \left[ \frac{p(x)}{J}, \frac{q_f(x)}{J} \right] - \lambda' \underbrace{d(y)}_{\text{" } d(x)} = 0$$

$$\text{So, } \underbrace{g \left[ \frac{p(x)}{J}, \frac{q_f(x)}{J} \right]}_{\text{since } J \text{ is arbitrary, it has to vanish from LHS:}} = \underbrace{(\lambda - \lambda') d(x)}_{\text{const}} + g[p(x), q_f(x)]$$

$$g \left[ \frac{p(x)}{J}, \frac{q_f(x)}{J} \right] = g \left[ \frac{p(x)}{q_f(x)} \right]$$

$$\downarrow \\ f[p(x), q_f(x)] = p(x) h \left[ \frac{p(x)}{q_f(x)} \right] + J[q_f(x)]$$

$$\text{Indeed, } \frac{\partial f[p(x), q_f(x)]}{\partial p(x)} = \underbrace{h \left[ \frac{p(x)}{q_f(x)} \right]}_{h \left[ \frac{p(x)}{q_f(x)} \right]} + h' \left[ \frac{p(x)}{q_f(x)} \right] \frac{p(x)}{q_f(x)}$$

$h \left[ \frac{p(x)}{q_f(x)} \right]$  is arbitrary thus far

$$g \left[ \frac{p(x)}{q_f(x)} \right]$$

Finally, Axiom 4 (system indep.):

Consider two indep. systems described by  $x_1$  &  $x_2$ : (2 constraints)

$$\int \mathcal{D}[x] d_k(x_k) p(x_1, x_2) = A_k \quad k=1, 2$$

$$\text{Define } r(x) = \frac{p(x)}{d_k(x)}, \quad x = \{x_1, x_2\}$$

Now,  $H = \int \mathcal{D}[x] p(x) h(r(x))$  gives

$$\begin{aligned} \frac{\delta}{\delta p(x)} \left[ H - \lambda_1 \int \mathcal{D}[x] p(x_1, x_2) d_1(x_1) - \right. \\ \left. - \lambda_2 \int \mathcal{D}[x] p(x_1, x_2) d_2(x_2) \right] = \\ = h(r(x)) + r(x) h'(r(x)) - \lambda_1 d_1(x_1) - \\ - \lambda_2 d_2(x_2) \stackrel{\text{system indep.}}{=} h[r_1, r_2] + r_1 r_2 h'[r_1, r_2] - \\ - \lambda_1 d_1 - \lambda_2 d_2 = 0. \end{aligned}$$

$$\begin{aligned} \frac{\partial^2}{\partial x_1 \partial x_2} : \quad & \frac{\partial}{\partial x_1} \left\{ h'[r_1, r_2] (\cancel{r_1 r_2} + r_1' r_2) + \right. \\ & + r_1 r_2 h''[r_1, r_2] (\cancel{r_1 r_2} + r_1' r_2') + \\ & + \cancel{r_1 r_2 h'[r_1, r_2]} + r_1 r_2' h'[r_1, r_2] - \\ & \left. - \lambda_2 d_2' \right\} = \\ = & h''[r_1, r_2] (\cancel{r_1 r_2}) (\cancel{r_2' r_1}) + \\ & + \cancel{h'[r_1, r_2]} (\cancel{r_2' r_1'}) + h'''[r_1, r_2] (r_1/r_2) (r_1^2 r_2 r_2') + \\ & + \cancel{h''[r_1, r_2]} (\cancel{2r_1 r_1' r_2 r_2'}) + \cancel{r_1' r_2' h'[r_1, r_2]} + \\ & + \cancel{r_1 r_2' h''[r_1, r_2]} (\cancel{r_1' r_2}) = \\ = & r_1' r_2' \{ 4r_1 r_2 h'' + 2h' + r_1^2 r_2^2 h''' \} = 0, \quad \text{or} \end{aligned}$$

$$4r^2 h'' + 2h' + r^2 h''' = 0$$

"r<sub>1</sub>r<sub>2</sub> (system indep.)

$$h(r) = -K \log(r) + B + C/r$$

↑ scale factor, >0      ↑ some const  
 (just a const in H)

$$\begin{aligned} \text{So, } H &= \int D[x] p(x) \left[ -K \log(r_{\text{fix}}) + B + C \frac{q_f(x)}{p(x)} \right] = \\ &= -K \int D[x] p(x) \log \left[ \frac{p(x)}{q_f(x)} \right] + \\ &\quad + C \int D[x] q_f(x) \underset{\substack{\text{also an irrelev.} \\ \text{const}}}{=} -K \int D[x] p(x) \log \left[ \frac{p(x)}{q_f(x)} \right]. \end{aligned}$$

Note that

$$H = -K \int D[x] p_1 p_2 \left[ \log \frac{p_1}{q_{f1}} + \log \frac{p_2}{q_{f2}} \right] =$$

$$= -K \left[ \int Dx_1 p_1 \log \frac{p_1}{q_{f1}} + \int Dx_2 p_2 \log \frac{p_2}{q_{f2}} \right] =$$

$$= H_1 + H_2 \underset{=}{\equiv}, \text{ as expected}$$

Note that  $H$  is a KL distance between  $p(x)$  &  $q_f(x)$

Maximize  $H \rightarrow$  minimize distance between  $p(x)$  &  $q_f(x)$  given the constraints

## Types of constraints

higher moments, combinations of moments?

$$S \int dx p(x) a(x)$$

Nonlinear in  $p(x)$  constraints?

Axiom 1: unique max of  $H$  under constraints  $\rightarrow$  no constraints that would mess up the convexity of

$$\int dx p(x) \log p(x)$$

What should  $a(x)$  be? Could be higher moments, but not always useful...

Thermodynamics: a system is in contact with a large bath

System + bath = ~~closed~~ closed

$$\text{Then } F = - \sum_{\substack{i,a \\ \uparrow \\ \text{System}}} p_{ia} \log p_{ia} + \lambda \left( \sum_{i,a} p_{ia} - 1 \right) + \lambda \sum_{\substack{i,a \\ \uparrow \\ \text{const}}} p_{ia} (1 - \delta_{E_i+E_a, E_{\text{tot}}})$$

$p_{ia} = 0 \text{ if } E_i + E_a \neq E_{\text{tot}}$

Note that  $p_i = \sum_a p_{ia}$ ,  $p(a|i) = \frac{p_{ia}}{p_i} \leftarrow p_{ia} = p(a|i)p_i$  [indep. not assumed]

$$\sum_i p_i = 1, \quad \sum_a p(a|i) = 1$$

$$\begin{aligned} \text{Then } - \sum_{i,a} p_{ia} \log p_{ia} &= \sum_{i,a} p(a|i)p_i [\log p(a|i) + \\ &\quad + \log p_i] = - \sum_i p_i \log p_i - \\ &\quad - \sum_{i,a} p(a|i)p_i \log p(a|i). \end{aligned}$$

↗ =

$$\text{So, } F_{\text{new}} = - \sum_i p_i \log p_i - \sum_{i,a} p(a|i) p_i \log p(a|i) +$$

$$+ \sum_i J_i \left[ \sum_a p(a|i) - 1 \right] \delta_{E_i + E_a, E_{\text{tot}}} + \lambda \sum_{i,a} p(a|i) p_i [1 - \delta_{E_i + E_a, E_{\text{tot}}}] .$$

$\delta F_{\text{new}}$  wrt  $p(a|i)$ ,  $J_i$ ,  $\lambda$ :

$$\left. \sum_{i,a} p(a|i) p_i [1 - \delta_{E_i + E_a, E_{\text{tot}}}] \right|_{\substack{(p_i \& \lambda \text{ given}) \\ \text{open system vars}}} = 0,$$

$$\sum_a p(a|i) = 1, \text{ if } E_i + E_a = E_{\text{tot}}$$

$$- p_i \log p(a|i) - p_i + J_i + \lambda p_i [1 - \delta_{E_i + E_a, E_{\text{tot}}}] = 0$$

$$\downarrow$$

$$\log p(a|i) = \begin{cases} \frac{J_i}{p_i} - 1, & E_i + E_a = E_{\text{tot}} \\ \frac{J_i}{p_i} + \lambda - 1, & E_i + E_a \neq E_{\text{tot}} \end{cases}$$

$$\sum_{i,a} p(a|i) p_i = 0, \quad \text{or } p(a|i) = 0, \quad \forall a$$

$$E_i + E_a \neq E_{\text{tot}} : \quad \text{consistent with } \lambda = -\infty \quad \# \text{ microstates at } E_a \neq E_{\text{tot}} - E_i = 0$$

$$\sum_{i,a} e^{\frac{J_i}{p_i} + \lambda - 1} p_i = 0 \quad \Rightarrow \quad (\sum_a 1) (\sum_i e^{\frac{J_i}{p_i}} p_i) = 0$$

$$\sum_a e^{\frac{J_i}{p_i} + \lambda - 1} = 1 \quad \Rightarrow \quad (\sum_a 1) = e^{1 + \lambda - \sum_i \frac{J_i}{p_i}} \Rightarrow \lambda = +\infty$$

$$E_i + E_a = E_{\text{tot}} : \quad \sum_a 1 = \sqrt{2(E_{\text{tot}} - E_i)} \quad \# \text{ bath microstates w/ energy } E_a = E_{\text{tot}} - E_i$$

$$\sum_a e^{\frac{J_i}{p_i} - 1} = 1 \quad \Rightarrow \quad p(a|i) = \frac{1}{\sqrt{2(E_{\text{tot}} - E_i)}} \quad \text{indep. of } a$$

$$\text{So, } p(a|i) = \frac{\delta_{E_a, E_{\text{tot}} - E_i}}{\sqrt{2(E_{\text{tot}} - E_i)}} \quad \equiv$$

Now,

$$F_{\text{new}} = - \sum_i p_i \log p_i + \lambda \left( \sum_i p_i - 1 \right) -$$
$$- \sum_{i,a} p_i \underbrace{\frac{\delta E_a, E_{\text{tot}} - E_i}{\mathcal{R}(E_{\text{tot}} - E_i)} \log \frac{\delta E_a, E_{\text{tot}} - E_i}{\mathcal{R}(E_{\text{tot}} - E_i)}}_{\text{"0 unless } E_a + E_i = E_{\text{tot}}\text{ and } \sum_a \frac{1}{\mathcal{R}(E_{\text{tot}} - E_i)} = 1\text{"}} =$$
$$= - \sum_i p_i \log p_i + \lambda \left( \sum_i p_i - 1 \right) + \sum_i p_i \log \mathcal{R}(E_{\text{tot}} - E_i)$$

Large bath:  $E_{\text{tot}} \gg E_i$ ,

$$\log \mathcal{R}(E_{\text{tot}} - E_i) \approx \log \mathcal{R}(E_{\text{tot}}) - \beta E_i$$

So the constraint becomes

$$-\beta \sum_i p_i E_i, \text{ just as before}$$

Higher-order moments drop out when the bath is large  $\rightarrow$  1st order moments arise naturally.

Indeed, as  $N \rightarrow \infty$  in any system,

$$Q = \sum_E g(E) e^{-\beta E}$$

energy level degeneracy

fluct's  $\langle E^2 \rangle - \langle E \rangle^2$  become small as  $N \uparrow$ ,  
(as  $\frac{1}{N}$ )

only  $\langle E \rangle$  remains finite

Higher-order moments also vanish...

$\hookrightarrow$  In "nanothermodynamics", higher-order moments may play a role, but the MAXENT framework still holds.

MAXENT framework  $\hookrightarrow$  still holds.