

Large deviation theory:

- Sanov theorem - Jaynes Max Ent principle

n i.i.d w_i random variables with values in set Λ
with prob. $p_j \quad 1 \leq j \leq |\Lambda|$

Empirical vector $L_{n,j}(w) = \frac{1}{n} \sum_{i=1}^n \delta_{w_i,j}$

Gärtner-Ellis theorem for random vectors $k \in \mathbb{R}^{|\Lambda|}$

scaled cumulant generating function (k is a vector with $|\Lambda|$ components)

$$\lambda(k) = \lim_{n \rightarrow \infty} \frac{1}{n} \ln \langle \exp(n k \cdot L_n) \rangle$$

$$= \ln \sum_{j=1}^{|\Lambda|} p_j \exp(k_j)$$

$\lambda(k)$ analytic $\forall k \in \mathbb{R}^{|\Lambda|} \Rightarrow$

we can apply the G-E th \Rightarrow

L.D. principle holds for $L_n(w)$ with
rate function $I(\ell)$

$$I(\ell) = \sup_k [k \cdot \ell - \lambda(k)] \quad \nabla_k \lambda = \ell$$

$$\rightarrow = k^*(\ell) \cdot \ell - \lambda(k^*) \quad \text{with } k^* / \nabla \lambda = \ell$$
$$\nabla \lambda(k) = \frac{\sum_j p_j \exp(k_j) \delta_{k_j}}{\sum_e p_e \exp(k_e)} = \frac{p \exp(k)}{\sum_e p_e \exp(k_e)}$$

$$\nabla \lambda = \ell \rightarrow \frac{\rho_j \exp(\kappa_j^*)}{\sum_c \rho_c \exp(\kappa_c^*)} = \ell_j$$

$\exp[\lambda(\kappa_j^*)]$

$$\kappa_j^* = \ln(\ell_j / \rho_j) + \lambda(\kappa^*)$$

$$I(\ell) = \kappa^* \cdot \ell - \lambda(\kappa^*) = \sum_j \ell_j \ln(\ell_j / \rho_j) + \underbrace{(\sum_j \ell_j)}_1 \lambda(\kappa^*) - \lambda(\kappa^*)$$

$$I(\ell) = \sum_j \ell_j \ln(\ell_j / \rho_j) = D[\ell / \rho]$$

→ Jensen theorem

Easily generalized to a continuous "empirical vector"

$$L_m(x) = \frac{1}{m} \sum_{i=1}^m \delta(w_i - x) \quad x \in \mathbb{R}$$

w distributed according to $\rho(x)$

$$I(\mu) = D[\mu / \rho] = \int_{-\infty}^{+\infty} dx \mu(x) \ln \left[\frac{\mu(x)}{\rho(x)} \right]$$

Jaynes Max Ent. principle

m particles with single particle energies $\epsilon_1, \dots, \epsilon_m$

$$\text{Total energy } h_m = \frac{1}{m} \sum_{i=1}^m \epsilon_i \quad \left(\begin{array}{l} \text{energy} \\ \text{density} \end{array} \right)$$

(non interacting particles)

One-particle energy distribution (empirical vector)

$$L_n(\varepsilon) = \frac{1}{n} \sum_{i=1}^n \delta(\varepsilon_i - \varepsilon)$$

fraction of particles with given energy

$$\tilde{h}(L_n) = \int \varepsilon L_n(\varepsilon) d\varepsilon$$

"contractive"

$L_n(\varepsilon) \rightarrow$ "microstate"

max ent principle: $S(u) = \sup_{\mu | \tilde{h}(\mu) = u} [-I^n(\mu)]$

Sanov theorem $\rightarrow I^n(\mu) = -\tilde{S}(\mu) + \ln |\Lambda|$

$$\tilde{S}(\mu) = - \int_{\Lambda} d\varepsilon \mu(\varepsilon) \ln(\mu(\varepsilon))$$

Shannon entropy of $\mu(\varepsilon)$

uniform prior distribution for $\mu(\varepsilon)$

$$S(u) = \sup_{\mu: \tilde{h}(\mu) = u} [\tilde{S}(\mu)] - \ln |\Lambda|$$

JAYNES
MAX ENT
Principle