

Sharing of probabilistic information of Bayesian agents

Bayesian agents

Bayesian agent

= set of random variables + joint domain + joint pdf

S agents

common set of variables

$(X_1, \dots, X_K) = \mathbf{X}$

common finite domain

$\{\mathbf{x}_1, \dots, \mathbf{x}_N\}$

pmfs (probability vectors)

q_1, \dots, q_S

true pmf

p

Supra Bayesian approach

scalar vectors q_1, \dots, q_S, p vs. random vectors $\mathbf{q}_1, \dots, \mathbf{q}_S, \mathbf{p}$

scalar matrix Q vs. random matrix \mathbf{Q}

\mathbf{p} (minus the last element) is continuously distributed
cpdf $t_{\mathbf{p}|\mathbf{Q}}$ given $\mathbf{Q} = Q$

estimate \mathbf{p}

minimize $E [K(\mathbf{p}||\hat{\mathbf{p}}) | \mathbf{Q} = Q]$

solution $\hat{\mathbf{p}} = E_{t_{\mathbf{p}|\mathbf{Q}}} [\mathbf{p} | \mathbf{Q} = Q]$

Construction of $t_{p|Q}$

principle of maximum entropy

principle of indifference wouldn't suffice

constraints $E_{t_{p|Q}} [K (q_s || p) | \mathbf{Q} = Q] = I_s$

set of simultaneous implicit equations

Dirichlet distribution

$$p' = w_0 q_0 + w_1 q_1 + \dots + w_s q_s$$

Nuance

$K(q_s \parallel \mathbf{p})$ vs $K(\mathbf{p} \parallel q_s)$

Kalenkovich vs Sechkarova

evaluation vs. compromise

union of supports vs. intersection

Extensions

- different supports
 - union
- generalized moments instead of pmfs
 - maximum entropy principle
- different supports + conditional marginal pmfs
 - (X_1, X_2, X_3) , first agent - $q_{X_2|X_1}$
 - extend q_1
 - minimize $E [K(q_1 || \mathbf{p}) | \mathbf{Q} = Q]$
 - cpmf calculated from $q_1 = q_{X_2|X_1}$
 - $q_1 = p'_{X_3|X_2, X_1} * q_{X_2|X_1} * p'_{X_1}$ on the *support* of $q_{X_2|X_1}$
 - $q_1 = p'$ elsewhere

Continuous case

S agents

common set of variables

common bounded domain

pdfs

true pdf

$$(X_1, \dots, X_K) = \mathbf{X}$$

A

$$g_1, \dots, g_S$$

g

random counterparts $\mathbf{g}_1, \dots, \mathbf{g}_S, \mathbf{g}, \mathbf{G}$

the most entropic random process?

From continuous to discrete

the most entropic random process?

partition (A_1, \dots, A_N) of A

$$\mathbf{g} = I_{A_1} * \mathbf{p}_1 / B(A_1) + \dots + I_{A_N} * \mathbf{p}_N / B(A_N), \mathbf{p}, \rho$$

$$t_{\mathbf{p}|Q}, E_{t_{\mathbf{p}|Q}} [K (\mathbf{g}_s \parallel \mathbf{g}) \mid \mathbf{G} = \mathbf{G}] = k_s$$

arbitrary synthetic support $\{\mathbf{x}_1, \dots, \mathbf{x}_N\}$

$$q_s(\mathbf{x}_n) = \text{Int} (q_s(\mathbf{x}) d\mathbf{x}, \mathbf{x} \text{ in } A_n)$$

$$I_s = k_s + \text{Sum} (B(A_n) q_s(\mathbf{x}_n), n = 1, \dots, N)$$

What is missing.

- I_s ? where do they come from?
- better method for merging pdfs
- other, less important things