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ENTROPY X ... finite set ; μ ... measure on X

$$H(\mu) := - \sum_{x \in X} \log(\mu(x)) \cdot \mu(x)$$

RELATIVE ENTROPY (X, μ) ; (X, ν) ... μ, ν two measures on X

$$D(\mu \parallel \nu) := \sum_{x \in X} [\log \mu(x) - \log \nu(x)] \mu(x)$$

- where
 - summand is 0 whenever $\mu(x) = 0$
 - $\nu(x) = 0$ implies $\mu(x) = 0$

Measure preserving map : $\Psi: X \rightarrow Y$ is m.p. if $\mu(\Psi^{-1}(y)) = \nu(y)$

Joint factor $(X_i, \mu_i)_{i=1}^3$... finite probability spaces

Assume that $\begin{cases} \Psi_1: X_1 \rightarrow X_3 \\ \Psi_2: X_2 \rightarrow X_3 \end{cases}$ are measure preserving maps.

Then X_3 is a joint factor of (X_1, μ_1) and (X_2, μ_2)

Coupling of (X_1, μ_1) , and (X_2, μ_2) over (X_3, μ_3)

$$X_4 := \{(x_1, x_2) \in X_1 \times X_2 : \Psi_1(x_1) = \Psi_2(x_2), \mu_3(\Psi_1(x_1)) \neq 0\}.$$

The measure μ is called a coupling.. if the projections $\Pi_i: X_4 \rightarrow X_i$

are measure preserving on (X_4, μ) .

Let μ_n be the measure on X_n defined by

$$\mu_n((x_1, x_2)) := \frac{\mu_1(x_1) \cdot \mu_2(x_2)}{\mu_3(\Psi_1(x_1))}$$

Then it is clear that projections are measure preserving

Lemma: Among couplings, the conditionally independent coupling minimizes the relative entropy with respect to a conditionally independent coupling.

PIE-type expression $D(\mu_4 \parallel \nu_4) = D(\mu_1 \parallel \nu_1) + D(\mu_2 \parallel \nu_2) - D(\mu_3 \parallel \nu_3)$

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Vertex factor

μ ... prob. distribution on $\text{Hom}(H, G)$

$\beta: S \rightarrow V(H)$ be an injective map

Map $\phi \mapsto \phi \circ \beta$ [where $\phi \in \text{Hom}(H, G)$]

defines a factor of $(\text{Hom}(H, G), \mu)$.

$S \subseteq V(H)$

Denote this factor by $(V(G)^S, \mu|_S)$ and call it vertex factor

by $\mu|_S$ denote $\mu|_{\beta}$ with $\beta: S \rightarrow S$ identity

Coupling

$(\text{Hom}(H_1, G), \mu_1), (\text{Hom}(H_2, G), \mu_2)$... prob. spaces

$\{\beta_i: [n] \rightarrow V(H_i)\}_{i=1}^2$ such that $\mu_3 = \mu_1|_{\beta_1} = \mu_2|_{\beta_2}$

denote by $C(\mu_1, \mu_2, \beta_1, \beta_2)$ the conditionally

independent coupling of μ_1 and μ_2 over μ_3 .

Probability

Scheme:

of a graph H is a function f on the set of finite graphs whose value $f(G)$ is the probability distribution on $\text{Hom}(H, G)$

We say that H is the frame of the scheme f .

Joint factor

f_i ... prob. scheme for H_i

$\beta_i: [n] \rightarrow V(H_i)$... two labelings such that $\overbrace{f_1(G)|_{\beta_1}}^{HG} = f_2(G)|_{\beta_2}$

β_1 and β_2 define a joint vertex factor of f_1 and f_2 .

Coupling

The conditionally independent coupling $g = C(f_1, f_2, \beta_1, \beta_2)$

of f_1 and f_2 is the function g whose value on

G is $C(f_1(G), f_2(G), \beta_1, \beta_2)$

is the graph obtained by the disjoint union of H_1, H_2 and identifying the vertices with the same label.

The frame of g

A_1

Let A_1 denote the smallest set of prob. schemes containing the scheme $G \rightarrow \mathcal{T}(e, G)$ and closed with respect to conditionally independent couplings over independent vertex sets.

Lemma!

Every element in A_1 is a family of witness measures.