

Holevo Bound & Relative Entropy

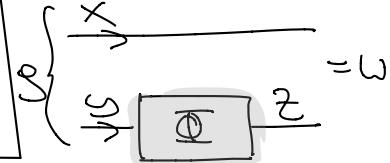
Last week: Entropy, Mutual info, properties

Recall: Strong subadditivity: $\forall \rho_{ABC} \in D(A \otimes B \otimes C)$:

$$I(A:B) \leq I(A:BC) \quad \text{i.e. } H(AB) + H(BC) \geq H(B) + H(ABC)$$

Data Processing inequality (DPI): $\forall g_{xy} \in D(x \otimes y), \Phi \in \mathcal{C}(y, z)$:

$$I(X:Z)_\omega \leq I(X:Y)_g \text{ for } \omega_{xz} = (\mathcal{I}_X \otimes \Phi)[g_{xy}]$$



* generalizes SSA: $X=A, Y=BC, Z=IB, \Phi=\text{tr}_C$

* implied by SSA: choose Stinespring isometry $V: Y \rightarrow Z \otimes W$ for Φ :

$$\omega_{xzw} = (\mathcal{I}_X \otimes V)g_{xy}(\mathcal{I}_X \otimes V^*) \text{ extends } \omega_{xz}$$

$$\& \quad I(X:Y)_g \stackrel{\text{P}}{\equiv} I(X:Z)_w \stackrel{\text{SSA}}{\geq} I(X:Z)_\omega$$

□

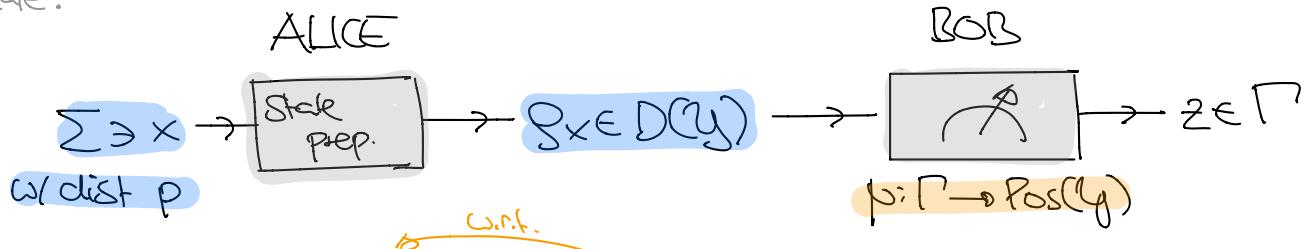
Recall: Ensembles $\{p_x, g_x\}$ can be represented by cq-states

$$g_{xy} = \sum_x p_x |x\rangle \langle x| \otimes g_x \in D(x \otimes y)$$

Holevo χ -quantity: $\chi(\{p_x, g_x\}) = I(X:Y) = H(\sum_x p_x g_x) - \sum_x p_x H(g_x)$

* $\chi \leq H(X)$, with " $=$ " if $\{g_x : p_x > 0\}$ pairwise orthogonal image \rightsquigarrow ESET !

Why care?



How large can $I(X:Z)$ be? $p(x,z) = p_x \cdot \text{tr}[g_x p(z)]$

Maximum over all p is known as "accessible information" of ensemble.

Thm (Holevo): $I(X:Z) \leq \chi(\{p_x, g_x\}) = I(X:Y)$

Pf: Apply DPI to p_{xy} and $\Phi[\sigma] = \sum_z \text{tr}[\sigma p(z)] |z\rangle\langle z|$ \square

$$\left(\omega_{xz} = \sum_x p_x |x\rangle\langle x| \otimes \Phi[p_x] = \sum_{x,z} p(x,z) |x\rangle\langle x| \otimes |z\rangle\langle z| \right)$$

Interpretation?

can decode X from Z
(i.e. $X = f(Z)$)

ESET 7

$$\Leftrightarrow H(XZ) = H(Z) \Leftrightarrow I(X:Z) = H(X)$$

$$\begin{aligned} & p_x > 0 && \text{if } H(x) && \text{PSET} \\ & \Leftrightarrow && x = H(x) \\ & \{g_x\} && \text{orthog.} \end{aligned}$$

"To communicate n random bits (perfectly), need to send n qubits."

$$\dim Y \geq |\Sigma|$$

* quantitative bounds via Fano's inequality * change of perspective: $x \rightarrow \text{channel} \rightarrow y$

Relative Entropy

Relative entropy: For $p, q \in P(\Sigma)$, define

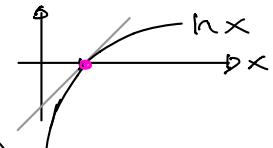
Consistent w/ $\log(\cdot + \epsilon)$ and $\epsilon \rightarrow 0$

$$D(p||q) = \begin{cases} \sum_x p(x) \log \frac{p(x)}{q(x)} & \text{if } \forall x: q(x) > 0 \Rightarrow p(x) > 0 \\ \infty & \text{otherwise} \end{cases}$$

NB: $p(x) \cdot \log q(x)$ clear if $p(x) > 0$, $= \begin{cases} 0 & \text{if } q(x) = p(x) = 0 \\ -\infty & \text{if } q(x) = 0, p(x) > 0 \end{cases}$

* $D(p||q) \geq 0, = 0 \text{ iff } p = q$ \leftarrow asymmetric distance measure

Pf: WLOG all $p(x) > 0$. Use $\ln(x) \leq x-1$, " $=$ " iff $x=1$.



$$D(p||q) = \sum_x p(x) \left(-\log \frac{q(x)}{p(x)} \right) \stackrel{\text{def}}{=} \frac{1}{\ln 2} \sum_x p(x) \left(1 - \frac{q(x)}{p(x)} \right) = 0$$

Why do we care?

"=" iff $q(x) = p(x) \quad \forall x$

\square

* $q(x) = \frac{1}{|\Sigma|}$: $D(p||q) = \log(|\Sigma| - H(p)) \geq 0, = 0 \text{ iff } p \text{ uniform}$ 😊

* $P_{xy} \in \mathcal{P}(\Sigma \times \Gamma)$, $q_{xy}(x,y) \equiv p_X(x)p_Y(y)$:

$$D(P_{xy}||q_{xy}) = \dots = I(X:Y)_p \geq 0, = 0 \text{ iff } P_{xy}(x,y) = p_X(x)p_Y(y) \quad \text{😊}$$

Quantum relative entropy: $S, \sigma \in D(\mathcal{A})$ Consistent w/ $\log(-\varepsilon I)$ and $\varepsilon \rightarrow 0$

$$D(g||\sigma) = \begin{cases} \text{tr}[g \cdot \log g] - \text{tr}[g \cdot \log \sigma] & \text{if } \text{ker}(\sigma) \subseteq \text{ker}(g) \\ \infty & \text{otherwise} \end{cases}$$
equiv:
 $\text{im}(g) \subseteq \text{im}(\sigma)$

* $g \cdot \log(\sigma)$ clear on $\text{ker}(\sigma)^{\perp}$, define as zero on $\text{ker}(\sigma)$ if $\text{ker}(\sigma) \subseteq \text{ker}(g)$

$\hookrightarrow D(g||\sigma) < \infty$ iff $\text{ker}(\sigma) \subseteq \text{ker}(g)$ e.g. $D(10x01 \parallel 1x11) = \infty$

* $S = \begin{pmatrix} p_{11} & \dots \\ \vdots & \ddots \end{pmatrix}, \sigma = \begin{pmatrix} q_{11} & \dots \\ \vdots & \ddots \end{pmatrix} : D(g||\sigma) = D(p||q)$

Monotonicity: $S, \sigma \in D(\mathcal{A})$, $\Phi \in C(\mathcal{A}, \mathcal{B})$

$$\Rightarrow D(g||\sigma) \geq D(\Phi[g] \parallel \Phi[\sigma])$$

"FUNDAMENTAL THEOREM
OF QIT"

Pf: HARD ↴

Klein's inequality: $D(g||\sigma) \geq 0, = 0$ iff $g = \sigma$

Pf: Let $p: \Gamma \rightarrow \text{Pos}(\mathcal{A})$ measurement & $\Phi \in C(\mathcal{A}, \mathcal{C}^{\Gamma})$ corresp. channel

$$\Phi[\omega] = \sum_{Y \in \Gamma} \text{tr}[\omega p(Y)] |Y\rangle\langle Y|$$

$$\begin{cases} p(Y) = \text{tr}[g p(Y)] \\ q(Y) = \text{tr}[\sigma p(Y)] \end{cases} \quad \begin{cases} \Phi[g] = \sum_Y p(Y) |Y\rangle\langle Y| \\ \Phi[\sigma] = \sum_Y q(Y) |Y\rangle\langle Y| \end{cases}$$

a) $D(g||\sigma) \geq D(\Phi[g] \parallel \Phi[\sigma]) = D(p||q) \geq 0$

b) $g = \sigma : D(g||\sigma) = 0 \quad \xrightarrow{\text{PSET}}$

c) $g \neq \sigma : \exists p : \|p - q\|_1, \|\Phi[g] - \Phi[\sigma]\|_1 > 0 \Rightarrow D(g||\sigma) \geq D(p||q) \stackrel{p \neq q}{\geq} 0 \quad \square$

~ PSET / ESET for further properties

Applications: Quick proofs of things we proved last time

ESET

* $D(g \parallel \frac{I_X}{d}) \stackrel{\text{ESET}}{=} \log d - H(g) \rightsquigarrow H(g) \stackrel{\text{klein}}{\leq} \log d, \stackrel{\text{ESET}}{=} \text{iff } g = \frac{I_X}{d} \quad \infty$

* Subadditivity: $D(g_{xy} \parallel g_x \otimes g_y) \stackrel{\text{ESET}}{=} I(X:Y)_{g_{xy}} \stackrel{\text{klein}}{\geq} 0, \stackrel{\text{ESET}}{<} 0 \text{ if } g_{xy} = g_x \otimes g_y \quad \infty$

* Strong subadditivity:

$$I(X:YZ) = D(\rho_{XYZ} \| \rho_X \otimes \rho_{YZ}) \stackrel{\text{Monotonicity}}{\geq} D(\rho_{XY} \| \rho_X \otimes \rho_Y) = I(X:Y) \quad \otimes$$

$\Phi = \rho_{YZ}$

* Basis measurement: $\rho \in D(\mathbb{C}^{\Sigma})$, $p(x) = \langle x | \rho | x \rangle$, $\sigma = \sum_x p(x) |x\rangle\langle x|$

$$\Rightarrow H(\rho) = H(\sigma) \geq H(\rho), \quad \text{iff } \rho = \sigma$$

Pf: $D(\rho \| \sigma) = -H(\rho) - \text{tr}[\rho \cdot \log(\sigma)] \stackrel{\text{iff}}{=} -H(\rho) + H(\rho) \rightsquigarrow \text{KLEIN. } \square$

$$\log(\sigma) = \sum_x \log(\langle x | \rho | x \rangle) |x\rangle\langle x|$$