

# TODAY: DATA PROCESSING FOR QUANTUM RELATIVE ENTROPY.

Quantum channel  $\mathcal{E}$

$$D(\mathcal{E}(\rho) \parallel \mathcal{E}(\sigma)) \leq D(\rho \parallel \sigma).$$

- Plan:
- Show that  $(\rho, \sigma) \mapsto D(\rho \parallel \sigma)$  is **convex**
  - Convexity  $\Rightarrow$  data processing.
  - Implications.

Th (**Joint convexity** of quantum relative entropy)

The function  $(\rho, \sigma) \mapsto \text{Tr}(\rho \log \rho - \rho \log \sigma)$   
 $\text{Pos}(\mathcal{H}) \times \text{Pos}(\mathcal{H}) \rightarrow \mathbb{R} \cup \{+\infty\}$

is convex, i.e., for any  $p \in [0, 1]$ ,  $\rho_0, \rho_1, \sigma_0, \sigma_1 \in \text{Pos}(A)$ :

$$D(p\rho_0 + (1-p)\rho_1 \parallel p\sigma_0 + (1-p)\sigma_1) \leq pD(\rho_0 \parallel \sigma_0) + (1-p)D(\rho_1 \parallel \sigma_1).$$

Moreover:

$$(*) \quad D(\rho \parallel \sigma) = \sup_{\substack{\{z_t\}_{t \geq 0} \\ z_t \in \mathcal{L}(\mathcal{H})}} \int_0^\infty \text{Tr} \left[ \rho \left( \frac{\mathbb{I} - z_t z_t^*}{t+1} - \frac{z_t z_t^*}{t} \right) - \sigma (\mathbb{I} + z_t^*)(\mathbb{I} + z_t) \right] dt$$

$\uparrow$   
 sup of linear functions

**linear** function in  $\rho, \sigma$

Rk: • Expression of  $D$  makes it manifestly convex as a sup of convex functions:

$$\sup_{f \in \mathcal{F}} f((1-p)\mu_0 + p\mu_1, (1-p)\sigma_0 + p\sigma_1) \stackrel{f \text{ linear}}{=} \sup_{f \in \mathcal{F}} (1-p)f(\mu_0, \sigma_0) + pf(\mu_1, \sigma_1) \\ \leq (1-p) \sup_{f \in \mathcal{F}} f(\mu_0, \sigma_0) + p \sup_{f \in \mathcal{F}} f(\mu_1, \sigma_1)$$

•  $\log = \ln$  here.

•  $\otimes$  also allows one to prove data processing easily.

Proof:  $\rho$  and  $\sigma$  do not commute so  
 $\log \rho - \log \sigma \neq \log \rho \sigma^{-1}$

Trick: write  $D(\rho \parallel \sigma)$  using one log.

$$\text{Tr}(A \cdot B) = \langle \bar{\Phi} | A \otimes B^T | \bar{\Phi} \rangle \quad |\bar{\Phi}\rangle = \sum_i |i\rangle \otimes |i\rangle$$

$\{|i\rangle\}$  basis of  $\mathcal{H}$ .

$$D(\rho \parallel \sigma) = \langle \bar{\Phi} | \rho \log \rho \otimes I - \rho \otimes \log(\sigma^T) | \bar{\Phi} \rangle.$$

$$= \langle \bar{\Phi} | (\rho \otimes I) (\log \rho \otimes I - I \otimes \log \sigma^T) | \bar{\Phi} \rangle$$

$$= \langle \bar{\Phi} | (\rho \otimes I) \log(\rho \otimes (\sigma^{-1})^T) | \bar{\Phi} \rangle. \quad \begin{array}{l} \log(A \otimes B) \\ = \log A \otimes I + I \otimes \log B \end{array}$$

$$\log(x) = \int_0^\infty \left( \frac{1}{t+1} - \frac{1}{t+x} \right) dt \quad \text{for } x > 0.$$

$$A \otimes I \log(A \otimes B^{-1}) = (A \otimes I) \int_0^\infty \frac{1}{(t+1)(I \otimes I)} - \frac{1}{tI \otimes I + A \otimes B^{-1}} dt$$

$$= \int_0^\infty \left( \frac{A \otimes I}{t+1} - \frac{1}{tA \otimes I + I \otimes B^{-1}} \right) dt$$

Parallel sum:  $(A : B) = \frac{1}{A^{-1} + B^{-1}}$  (operator harmonic mean)

Rk: Can define it for general  $A, B \in \text{Pos}(\mathcal{H})$ . For simplicity, here assume  $A, B$  invertible.

Proposition: (Variational expression for parallel sum)

For every  $x \in \mathcal{H}$ ,  $A, B \in \text{Pos}(\mathcal{H})$ . *linear in A, B.*

$$\langle x, (A:B)x \rangle = \inf \{ \langle y, Ay \rangle + \langle z, Bz \rangle : y+z=x \}$$

Proof:  $A:B = (A^{-1} + B^{-1})^{-1} = (B^{-1}(A+B)A^{-1})^{-1} = A(A+B)^{-1}B = (A+B-B)(A+B)^{-1}B$   
 $= B - B(A+B)^{-1}B.$

$$\begin{aligned} & \langle y, Ay \rangle + \langle x-y, B(x-y) \rangle - \langle x, (A:B)x \rangle \\ &= \langle x, Bx \rangle + \langle y, (A+B)y \rangle - 2\text{Re} \langle y, Bx \rangle - \langle x, (A:B)x \rangle \\ &= \langle x, B(A+B)^{-1}Bx \rangle + \langle y, (A+B)y \rangle - 2\text{Re} \langle y, Bx \rangle \\ &= \| (A+B)^{-\frac{1}{2}} Bx \|^2 + \| (A+B)^{\frac{1}{2}} y \|^2 - 2\text{Re} \langle (A+B)^{\frac{1}{2}} y, (A+B)^{-\frac{1}{2}} Bx \rangle \\ &= \| (A+B)^{-\frac{1}{2}} Bx - (A+B)^{\frac{1}{2}} y \|^2 \\ &\geq 0 \end{aligned}$$

with equality if  $y = (A+B)^{-1} Bx$  ■

Back to our expression

$$D(e||\sigma) = \langle \Phi | \int_0^\infty \left( \frac{e \otimes I}{t+1} - \left( \frac{e \otimes I}{t} \right) : (\mathbb{I} \otimes \sigma^T) \right) dt | \Phi \rangle$$

$$= \int_0^\infty \left( \langle \Phi | \frac{e \otimes I}{t+1} | \Phi \rangle - \langle \Phi | \left( \frac{e \otimes I}{t} \right) : (\mathbb{I} \otimes \sigma^T) | \Phi \rangle \right) dt. \quad (**)$$

$$\bullet \langle \Phi | \frac{e \otimes I}{t+1} | \Phi \rangle = \frac{1}{t+1} \text{Tr}(e \cdot \mathbb{I}^T) = \frac{1}{t+1} \text{Tr}(e).$$

$$\bullet \langle \Phi | \left( \frac{e \otimes I}{t} \right) : (\mathbb{I} \otimes \sigma^T) | \Phi \rangle = \inf_{|z\rangle \in \mathcal{H} \otimes \mathcal{H}} \langle z, \frac{e \otimes I}{t} z \rangle + \langle \Phi - \langle z | \mathbb{I} \otimes \sigma^T (| \Phi \rangle - | z \rangle)$$

Let us write  $|z\rangle$  in the fixed basis  $\{|i\rangle\}$ .  $|z\rangle = \sum_{ij} z_{ij} |i\rangle \otimes |j\rangle$

$$\bullet \frac{1}{t} \langle z | e^{\otimes t} | z \rangle = \frac{1}{t} \sum_{\substack{ij \\ i'j'}} \langle i | \langle j | \bar{z}_{ij} (e^{\otimes t}) z_{i'j'} | i' \rangle | j' \rangle$$

$$= \frac{1}{t} \sum_{\substack{ii' \\ j}} \bar{z}_{ij} \langle i | e | i' \rangle z_{i'j}$$

$$= \frac{1}{t} \text{Tr}(e Z Z^*) \quad \text{where } Z = \sum_{ij} z_{ij} |i\rangle \langle j|$$

$$Z^* = \sum_{ij} \bar{z}_{ij} |j\rangle \langle i|$$

$$\bullet \langle \Phi | I \otimes \sigma^T | \Phi \rangle = \text{Tr}(\sigma)$$

$$\bullet \langle z, I \otimes \sigma^T z \rangle = \sum_{\substack{ij \\ i'j'}} \langle i | \langle j | \bar{z}_{ij} (I \otimes \sigma^T) z_{i'j'} | i' \rangle | j' \rangle$$

$$= \sum_{ijj'} \bar{z}_{ij} z_{ij'} \langle j | \sigma^T | j' \rangle$$

$$= \text{Tr}(\sigma Z^* Z)$$

$$Z^* Z = \sum_{jj'} \bar{z}_{ij} z_{ij'} |j\rangle \langle j'|$$

$$\bullet \langle \Phi, I \otimes \sigma^T z \rangle = \sum_{ij} \bar{z}_{ij} \langle i | \sigma^T | j \rangle = \text{Tr}(\sigma Z)$$

$$\bullet \langle z, I \otimes \sigma^T \Phi \rangle = \sum_{ij} \bar{z}_{ij} \langle j | \sigma^T | i \rangle = \text{Tr}(\sigma Z^*)$$

Back to  $(**)$

$$\langle \Phi | \frac{e^{\otimes t+1}}{t+1} | \Phi \rangle - \langle \Phi | \left( \frac{e^{\otimes t}}{t} : I \otimes \sigma^T \right) | \Phi \rangle$$

$$= \sup_{Z_t} \left\{ \frac{1}{t+1} \text{Tr}(e) - \frac{1}{t} \text{Tr}(e Z Z^*) - \text{Tr}(\sigma (I + Z + Z^* + Z^* Z)) \right\} \blacksquare$$

From joint convexity to data processing (Rather generic, works for many divergences).

Th (Data processing inequality for the quantum relative entropy).

$$\left[ \begin{array}{l} \rho, \sigma \in \text{Pos}(A), \mathcal{E} \text{ quantum channel } L(A) \rightarrow L(B) \\ D(\mathcal{E}(\rho) \parallel \mathcal{E}(\sigma)) \leq D(\rho \parallel \sigma) \end{array} \right.$$

Proof: Consider a Stinespring dilation for  $\mathcal{E}$

$$\mathcal{E}(S) = \text{Tr}_E(VSV^*) \text{ with } V \in L(A, B \otimes E), V^*V = I_A$$

$$\mathcal{E} = \text{Tr}_E \circ \mathcal{V} \text{ where } \mathcal{V}(S) = VSV^*.$$

Claim:  $D(\mathcal{V}(\rho) \parallel \mathcal{V}(\sigma)) = D(\rho \parallel \sigma)$ .

$$\begin{aligned} \text{Tr}(V\rho V^* \log V\rho V^* - V\rho V^* \log V\sigma V^*) &= \text{Tr}(V\rho \log \rho V^* - V\rho \log \sigma V^*) \\ &= \text{Tr}(\rho \log \rho - \rho \log \sigma) \end{aligned}$$

So it suffices to analyze  $\text{Tr}_E$  map.

Let  $\rho_{BE}, \sigma_{BE} \in \text{Pos}(B \otimes E)$

Want to transform

$$\begin{array}{l} \rho_{BE} \longrightarrow \rho_B \otimes \frac{I}{\dim E} \\ \sigma_{BE} \longrightarrow \sigma_B \otimes \frac{I}{\dim E} \end{array}$$

$$D(\rho_{BE} \parallel \sigma_{BE}), \quad D\left(\rho_B \otimes \frac{I_E}{\dim E} \parallel \sigma_B \otimes \frac{I_E}{\dim E}\right) = D(\rho_B \parallel \sigma_B).$$

Consider generalized Pauli operators  $d = \dim E$ ,  $[d] = \{0, \dots, d-1\}$

$$X_E |k\rangle = |k+1 \pmod d\rangle, \quad Z_E |k\rangle = e^{\frac{2\pi i k}{d}} |k\rangle$$

Consider the channel  $\mathcal{Q}(S) = \frac{1}{d^2} \sum_{\ell, m \in [d]} X_E^\ell Z_E^m S Z_E^m X_E^\ell$

Then 
$$D(|h \times h'|) = \frac{1}{d^2} \sum_{\ell, m \in [d]} X_E^{\ell} e^{\frac{2\pi i(\ell-h')m}{d}} |h \times h'| X_E^{\ell}$$

$$= \begin{cases} 0 & \text{if } h \neq h' \\ \frac{1}{d} \sum_{\ell} X_E^{\ell} |h \times h| X_E^{\ell} = \frac{\mathbb{I}}{d} & \text{if } h = h' \end{cases}$$

So

$$(I_B \otimes D)(\rho_{BE}) = \rho_B \otimes \frac{I_E}{d}$$

$$(I_B \otimes D)(\sigma_{BE}) = \sigma_B \otimes \frac{I_E}{d}$$

Now use joint convexity:

$$D\left(\overbrace{\frac{1}{d^2} \sum_{\ell, m} X_E^{\ell} Z_E^m \rho_{BE} X_E^{\ell} Z_E^m}^{\rho_B \otimes \frac{I}{d}} \parallel \overbrace{\frac{1}{d^2} \sum_{\ell, m} X_E^{\ell} Z_E^m \sigma_{BE} X_E^{\ell} Z_E^m}^{\sigma_B \otimes \frac{I}{d}}\right)$$

$$\leq \frac{1}{d^2} \sum_{\ell, m} D\left(X_E^{\ell} Z_E^m \rho_{BE} X_E^{\ell} Z_E^m \parallel X_E^{\ell} Z_E^m \sigma_{BE} Z_E^m X_E^{\ell}\right)$$

$$\underbrace{\hspace{10em}}_{D(\rho_{BE} \parallel \sigma_{BE})}$$

Concludes the proof.  $\square$

# Direct proof of data processing using $*$

$$D(\rho \parallel \sigma) = \sup_{\{Z_t\}_{t \geq 0}, Z_t \in \mathcal{L}(\mathcal{H})} \int_0^\infty \text{Tr} \left[ e \left( \frac{I}{t+1} - \frac{Z_t Z_t^*}{t} \right) - \sigma (I + Z_t^*)(I + Z_t) \right] dt$$

$$D(\mathcal{E}(\rho) \parallel \mathcal{E}(\sigma)) = \sup_{\{Z_t\}_{t \geq 0}, Z_t \in \mathcal{L}(\mathcal{H})} \int_0^\infty \text{Tr} \left[ e \left[ \frac{\mathcal{E}(I)}{t+1} - \frac{\mathcal{E}(Z_t Z_t^*)}{t} \right] - \sigma \left[ \mathcal{E}(I) + \mathcal{E}(Z_t^*) + \mathcal{E}(Z_t) + \mathcal{E}(Z_t^* Z_t) \right] \right] dt$$

Claim: For a unital completely positive map  $\mathcal{F}$ ,

$$\mathcal{F}(Z Z^*) \geq [\mathcal{F}(Z)]^* \cdot \mathcal{F}(Z) \quad [\text{Schwarz map}]$$

Indeed  $T := |0\rangle\langle 0| \otimes I + |0\rangle\langle 1| \otimes Z + |1\rangle\langle 0| \otimes Z^* + |1\rangle\langle 1| \otimes Z^* Z \geq 0$

So  $(I \otimes \mathcal{F})(T) \geq 0$  i.e.,  $\begin{bmatrix} \mathcal{F}(Z Z^*) & \mathcal{F}(Z^*) \\ \mathcal{F}(Z) & I \end{bmatrix} \geq 0$

$$\begin{matrix} \uparrow \\ \mathcal{F}(Z Z^*) \geq \mathcal{F}(Z)^* \mathcal{F}(Z) \end{matrix}$$

So letting  $Y_t = \mathcal{E}^*(Z_t)$

$$\begin{aligned} D(\mathcal{E}(\rho) \parallel \mathcal{E}(\sigma)) &\leq \sup_{\{Y_t\}_{t \geq 0}, Y_t \in \mathcal{L}(\mathcal{H})} \int_0^\infty \text{Tr} \left[ e \left[ \frac{I}{t+1} - \frac{Y_t Y_t^*}{t} \right] - \sigma \left[ I + Y_t^* + Y_t + Y_t^* Y_t \right] \right] dt \\ &= D(\rho \parallel \sigma). \end{aligned}$$



# Consequences

\* Strong subadditivity:

$$H(A|C)_\rho + H(B|C)_\rho \leq H(AB|C)_\rho \quad (1)$$

$\Updownarrow$

$$H(ABC)_\rho \leq H(A|C)_\rho \quad (2)$$

$\Updownarrow$

$$I(A:B|C)_\rho \geq 0 \quad (3)$$

Proof: (2)  $H(ABC)_\rho = -D(\rho_{ABC} \| I_A \otimes \rho_{BC})$

$$H(A|B)_\rho = -D(\rho_{AB} \| I_A \otimes \rho_B)$$

Data processing:  $D(\rho_{AB} \| I_A \otimes \rho_B) \leq D(\rho_{ABC} \| I_A \otimes \rho_{BC})$

$$\begin{aligned} (2) \Rightarrow (1) \text{ Note that } H(ABC) &= H(ABC) - H(BC) \\ &= H(ABC) - H(C) - H(BC) + H(C) \\ &= H(ABC) - H(B|C) \end{aligned}$$

$$(2) \Rightarrow (3) \quad I(A:B|C)_\rho = H(A|C) - H(A|BC).$$