

From classical to quantum compression

Last week: Shannon entropy $H(p)$ = optimal rate for compressing a classical data source described by $p \in P(\Sigma)$.

Today: $g \in D(H)$? Let's first define entropy + then discuss compression.

Von Neumann Entropy of $g \in D(H)$:

$$\text{as usual: } g = \sum_i p_i |e_i\rangle\langle e_i| \quad f(g) := \sum_i f(p_i) |e_i\rangle\langle e_i|$$

$$H(g) = H(p) = -\sum_i p_i \log p_i = -\text{tr}[g \cdot \log g]$$

where p vector of eigenvalues of g (repeated acc. to multiplicity).

Properties:

- * $H(g) = H(VgV^*)$ for any isometry V

- * $0 \leq H(g) \leq \log \text{rk}(g) \leq \log d$, where $d := \dim H$

- * $= 0$ iff $\text{p.e. } = \log d$ iff $g = \frac{I}{d}$ (maximally mixed)

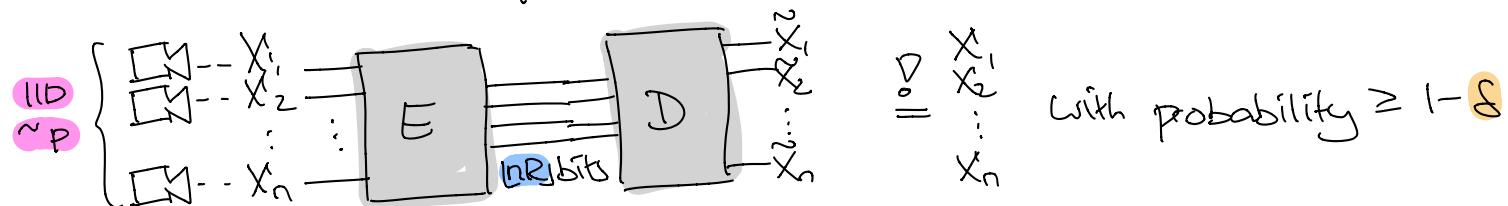
- * Continuous Pf: $g \propto \sigma \Rightarrow$ sorted eigenvalues of $g \approx$ those of σ (II)

- * Concave, i.e. $H\left(\sum_x p_x g_x\right) \geq \sum_x p_x H(g_x)$ HW

How to define the task of q. compression?

Motivation: Classical compression and correlations

Recall: (n, R, δ) -code for $p \in P(\Sigma)$:



$$\text{i.e. } \left\{ \begin{array}{l} E: \sum^n \rightarrow \{0,1\}^{nR} \\ D: \{0,1\}^{nR} \rightarrow \sum^n \end{array} \right\} \text{ s.t. } \Pr(\tilde{X}^n \neq X^n) \leq \delta$$

Shannon: $H(p)$ is optimal rate \rightarrow last week for precise statement

How about if $X_1, \dots, X_n \stackrel{\text{IID}}{\sim} p$, but correlated to another random variable Y ?

e.g. $Y = X_1$ or $Y = X_1 \oplus \dots \oplus X_n$ or even $Y = X^n$

$D(E(X^n))$

let P_{X^nY} , $P_{\tilde{X}^nY}$ joint distribution of (X^n, Y) and of (\tilde{X}^n, Y) , respectively

Lem: (D, E) is an (n, R, δ) -code iff $T(P_{X^nY}, P_{\tilde{X}^nY}) \leq \delta$ for all RV's (X^n, Y) s.t. $X^n \stackrel{\text{iid}}{\sim} p$

↑ trace dist of probability dist's

Sketch: (\Rightarrow) $T(P_{X^nY}, P_{\tilde{X}^nY})$

$$T(q, \tilde{q}) := \frac{1}{2} \sum_z |q(z) - \tilde{q}(z)|$$

$$\leq \Pr((X^n, Y) \neq (\tilde{X}^n, Y)) = \Pr(X^n \neq \tilde{X}^n) \leq \delta$$

(\Leftarrow) Choose $Y = X^n$? Then:

$$\Pr(D(E(X^n)) \neq X^n) = |\Pr(\tilde{X}^n \neq Y) - \Pr(X^n \neq Y)| \stackrel{\text{same}}{=} 0 \quad (\square)$$

→ PRACTICE

Thus: Correlations are preserved & this characterizes a reliable compression protocol.

We will take this as the def'n in the q. case!

Quantum compression

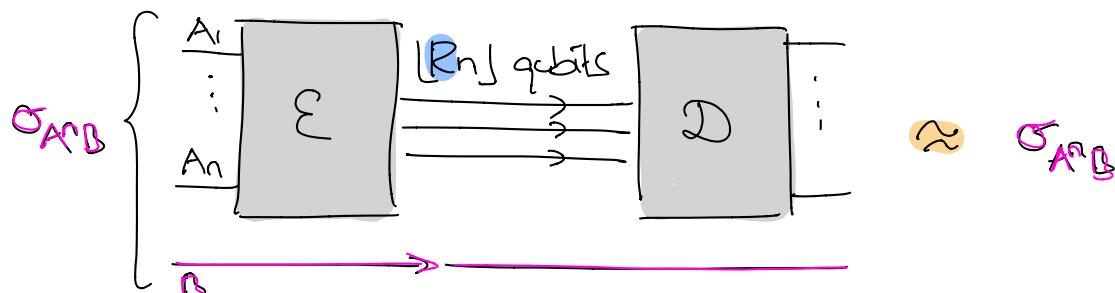
Def: (n, R, δ) -quantum code for $\mathcal{S} \in D(H_A)$: quantum channels

$$\mathcal{E} \in C(H_A^{\otimes n}, (\mathbb{C}^2)^{\otimes LR_n}) \text{ and } \mathcal{D} \in C((\mathbb{C}^2)^{\otimes LR_n}, H_A^{\otimes n})$$

s.t.

$$F((D \circ \mathcal{E} \otimes I_B)[\sigma_{A^nB}], \sigma_{A^nB}) \geq 1 - \delta$$

for all H_B , $\sigma_{A^nB} \in D(H_A^{\otimes n} \otimes H_B)$ with $\sigma_{A^n} = \mathcal{S}^{\otimes n}$

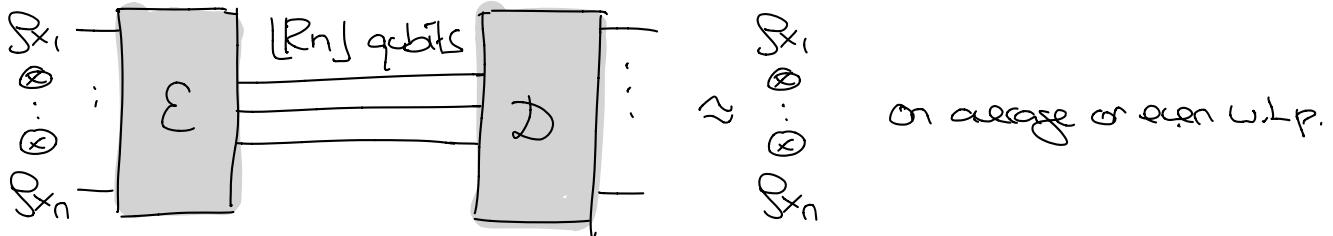


Some idea as
 $T \leq \delta$,
except that we
use fidelity

Plan: ① Relate to "ordinary" compression. ② Simplify definition
③ Prove analog of Shannon's theorem.

Surprising definition! How is this related to compression of a q. source?

WANT:



Choose (n, R, δ) -q. code for average output state $\bar{s} = \sum_x p(x) s_x$

Then:

$$\sum_{x^n} p(x_1) \dots p(x_n) F(s_{x_1} \otimes \dots \otimes s_{x_n}, D(E(s_{x_1} \otimes \dots \otimes s_{x_n}))) \geq 1 - \delta \quad \text{or}$$

Pf: Define $\sigma_{\text{Any}} = \sum_{x^n} p(x_1) \dots p(x_n) s_{x_1} \otimes \dots \otimes s_{x_n} \otimes |x^n\rangle\langle x^n| \in D(H_A^{\otimes n} \otimes H_Y)$,
 $H_A = \mathbb{C}^{\Sigma^n}$

$$\Rightarrow \text{LHS} \stackrel{\text{HWS}}{=} F(\sigma_{\text{Any}}, (D \circ E \otimes I_Y)[\sigma_{\text{Any}}]) \geq 1 - \delta \quad \square$$

Thus: Can compress Any source w/ average output state \bar{s} !

→ HW

WARNING: In general:

- $|\Sigma| \neq \dim H_A$
- s_x NOT pairwise orthogonal,
- $p(x)$ NOT eigenvalues of \bar{s}

Next, let's simplify the definition of a q. code...

Def: Channel fidelity of $T_A \in \mathcal{C}(H_A, H_A)$ and $s_A \in D(H_A)$:

$$F(\Phi_{AB}, s_A) = \inf \left\{ F(\Phi_{AB}, (T_A \otimes I_B)[\sigma_{AB}]) : H_B, \sigma_{AB} \text{ s.t. } \Phi_{AB}[\sigma_{AB}] = s_A \right\}$$

↳ Condition in (n, R, δ) -code can be written as $F(D \circ E, \bar{s}^{\otimes n}) \geq 1 - \delta$

Lem: Let $|\Psi_{AB}\rangle \in H_A \otimes H_B$ be any purification of s_A . Then:

$$F(T_A, s_A) = F(|\Psi_{AB}\rangle\langle\Psi_{AB}|, (T_A \otimes I_B)[|\Psi_{AB}\rangle\langle\Psi_{AB}|])$$

→ no more "for all σ_{AB} "!

Pf: Monotonicity → σ_{AB} p.c. Isometric invariance → Any purification is ok. □

Corr: If $T_A[\rho_{AB}] = \sum_i X_i \otimes I_B$ Kraus repr: $F(T_A[\rho_{AB}])^2 = \sum_i |\text{tr}[X_i \rho_{AB}]|^2$

Pf: Using the previous lemma:

$$\begin{aligned} F(T_A[\rho_{AB}])^2 &= \left\langle \Psi_{AB} \left| \sum_i (X_i \otimes I_B) \right| \Psi_{AB} \right\rangle \left\langle \Psi_{AB} \left| (X_i^\dagger \otimes I_B) \right| \Psi_{AB} \right\rangle \\ \text{Fiddling with one state part} &= \sum_i \left| \underbrace{\left\langle \Psi_{AB} \left| (X_i \otimes I_B) \right| \Psi_{AB} \right\rangle}_{} \right|^2 = \sum_i \left| \text{tr}[\rho_{AB} X_i] \right|^2 \\ &= \text{tr} \left[(\Psi_{AB}^\dagger \Psi_{AB}) \left(\sum_i (X_i \otimes I_B) \right) \right] \end{aligned} \quad \square$$

Schumacher's theorem & Typical Subspaces

Schumacher's Theorem: Let $0 < \delta < 1$:

- ① If $R > H(\rho)$: $\exists n_0: \forall n \geq n_0: \exists (n, R, \delta)$ -q code
- ② If $R < H(\rho)$: $\exists n_0: \forall n \geq n_0: \nexists (n, R, \delta)$ -q code

key tool:

Typical subspace: $S_{n,\varepsilon}(\rho) = \text{Span} \{ |y\rangle \otimes \dots \otimes |y\rangle : y^n \in T_{n,\varepsilon}(q) \}$

where $\rho = \sum_y q_y |y\rangle \langle y|$ eigen decomposition

$\Pi_{n,\varepsilon}$ = orthog. projection onto $S_{n,\varepsilon}(\rho)$ "typical projector"

Properties: ① Eigenvalues of $\Pi_{n,\varepsilon} \rho \otimes \Pi_{n,\varepsilon}$ are in $2^{-n(H(\rho) + \varepsilon)}$

① $\dim S_{n,\varepsilon}(\rho) = |T_{n,\varepsilon}(q)| \leq 2^{n(H(\rho) + \varepsilon)}$

② $\text{tr} [\Pi_{n,\varepsilon} \rho^{\otimes n}] \rightarrow 1$

$$= \sum_{y \in T_{n,\varepsilon}(q)} q(y_1) \dots q(y_n)$$

Pf: This follows directly from the corresponding properties of $T_{n,\varepsilon}(q)$. \square

Next week, we will use this to prove Schumacher's theorem.

→ **PRACTICE** & **LECTURE NOTES** for a sneak peak!