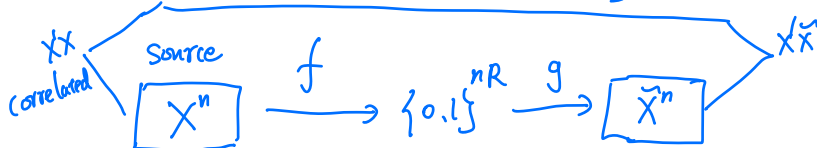


Recall classical source coding



$$\Sigma_n(R) = \inf P(X^n \neq \tilde{X}^n), \quad R \text{ rate of compression}$$

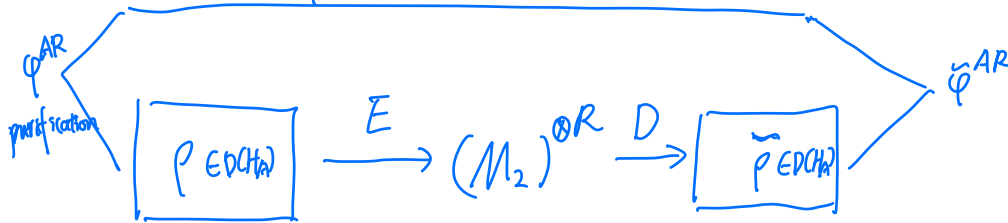
$$= \|P_{X^n X^n} - P_{X^n \tilde{X}^n}\|_{TV}$$

Shannon's Theorem

$$\lim_n \Sigma_n(R) = \begin{cases} 1 & \text{if } R > H(X) \\ 0 & \text{if } R < H(X) \end{cases}$$

$$P_{X^n X^n}(x, y) = \begin{cases} P_X^{\otimes n} & x = y \\ 0 & x \neq y \end{cases}$$

Quantum data compression

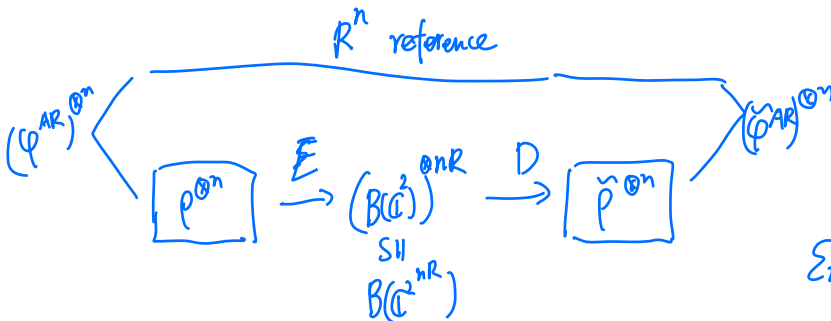


$$\Sigma(R) = \inf_{(E,D)} "Pr(\rho \neq \tilde{\rho})" = \inf_{(E,D)} \| \varphi^{AR} - E \circ D(\varphi^{AR}) \|_1$$

Classical case: $\|P - Q\|_{TV} = \frac{1}{2} \|P - Q\|_1 = \frac{1}{2} \sum_{x \in \mathcal{X}} |P(x) - Q(x)| = \inf_{X \sim P, \tilde{X} \sim Q} P(X \neq \tilde{X})$

Quantum case: $\|P - G\|_1 = \text{tr}(|P - G|) = \sup_{0 \leq P \leq I} \text{tr}((P - G)P) := \text{maximal guess probability.}$

I.I.d Setting



$$\Sigma_n(R) = \inf_{E,D} \| \varphi^{AR \otimes n} - E \circ D(\varphi^{AR \otimes n}) \|_1$$

Schumacher Compression (1994)

$$\lim_{n \rightarrow \infty} \varepsilon_n(R) = \begin{cases} 0 & \text{if } R > H(p) \\ 1 & \text{if } R < H(p) \end{cases}$$

Direct coding ($R > H(p)$)

$$p = \sum p_x |x\rangle\langle x|, \quad H(p) = H(P), \quad P_x = \{p_x\} \text{ over } x \in \mathcal{X}$$

Then if $R > H(p) = H(P)$, by Shannon's coding theorem, $\exists g_n, g_n^{-1}$ - detectable error

$$\boxed{P_X^n} \xrightarrow{f_n} \boxed{\{0,1\}^{nR}} \xrightarrow{g_n} \boxed{P_{X^n}^n}$$

$$\exists S_n \subseteq \mathcal{X}^n \text{ s.t. } g_n \circ f_n|_{S_n} = I, \quad \lim_{n \rightarrow \infty} P(X^n \neq \tilde{X}^n) = \lim_n P(S_n) = 0$$

Define partial isometry $V: (\mathbb{C}^{\mathcal{X}})^{\otimes n} \rightarrow (\mathbb{C}^2)^{\otimes nR}$

$$V|_{S_n} = \begin{cases} |f(x_1, \dots, x_n)\rangle & \text{if } x_1, \dots, x_n \in S_n \\ 0 & \text{otherwise} \end{cases}$$

$V^*V = \pi_V$ a projection

$$\text{Define } E: B(\mathbb{C}^{\mathcal{X}^n}) \rightarrow B(\mathbb{C}^{2^{nR}})$$

$$E(p) = V p V^* + \text{tr}((I - V^*V)p) |e\rangle\langle e| \text{ for some fixed } |e\rangle\langle e| \text{ for error}$$

$$D: B(\mathbb{C}^{2^{nR}}) \rightarrow B(\mathbb{C})^{\mathcal{X}^n}$$

$$D(p) = V^* p V + \text{tr}((I - VV^*)p) |e\rangle\langle e| \quad \text{tr}(\pi_V p) = \text{tr}(V p V^*) = P(S_n)$$

$$D \circ E(\varphi^{AR \otimes n}) = V^* V (\varphi^{AR \otimes n}) V^* V + \text{tr}((I - V^*V)p) D(|e\rangle\langle e|)$$

$$= \pi_V (\varphi^{AR \otimes n}) \pi_V + \text{tr}(\pi_V^c p) D(|e\rangle\langle e|)$$

$$\|(\varphi^{AR \otimes n}) - D \circ E(\varphi^{AR \otimes n})\|_1 \leq \|\varphi^{AR} - \pi_V (\varphi^{AR \otimes n}) \pi_V\|_1 + \text{tr}(\pi_V^c p) \|D(|e\rangle\langle e|)\|$$

$$P(S_n) \|D(|e\rangle\langle e|)\|$$

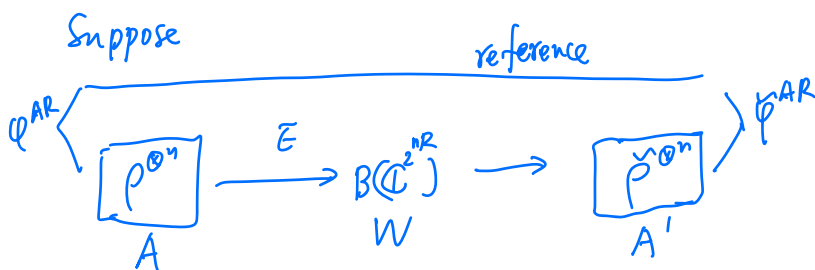
$$\leq 2\varepsilon_n + \varepsilon_n = 3\varepsilon_n \rightarrow 0$$

Weak converse:

Theorem (Fannes-Audenaert) For $\rho, \sigma \in \mathcal{B}(H)$ and $\varepsilon = \frac{1}{2} \|\rho - \sigma\|_1$,
 $|H(\rho) - H(\sigma)| \leq \varepsilon \log(\dim H - 1) + h(\varepsilon).$

Corollary: For $\rho^{AB}, \sigma^{AB} \in \mathcal{D}(H_{AB})$ and $\varepsilon = \frac{1}{2} \|\rho^{AB} - \sigma^{AB}\|_1$
 $|I(A:B)_\rho - I(A:B)_\sigma| \leq 3 \varepsilon \log(\dim H - 1) + 3h(\varepsilon)$

Pf: $I(A:B) = H(A) + H(B) - H(AB)$.



$$\begin{aligned} 2nR = 2 \log 2^{nR} &= 2 \log |W| \geq I(W, R^n) \geq I(A'^n, R^n) \stackrel{\varepsilon}{\geq} I(A'^n, R^n)_{\rho^{A'^n}} - 3\varepsilon \log |A'|^n - 3h(\varepsilon) \\ &\geq n I(A, R) - 3\varepsilon \log |A'|^n - 3h(\varepsilon) \\ &\geq 2nH(A) - 3\varepsilon n \log |A'| - 3h(\varepsilon) \end{aligned}$$

$$\Rightarrow \lim_{n \rightarrow \infty} \frac{1}{n} \log 2^{nR} \geq \frac{2}{3} \frac{H(A) - R}{\log |A'|} > 0.$$

If $R < H(A)$, $\exists \varepsilon$ s.t. $R + 3\varepsilon < H(A)$, then contradiction.