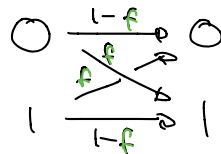


The Noisy Coding Theorem (§9-10)

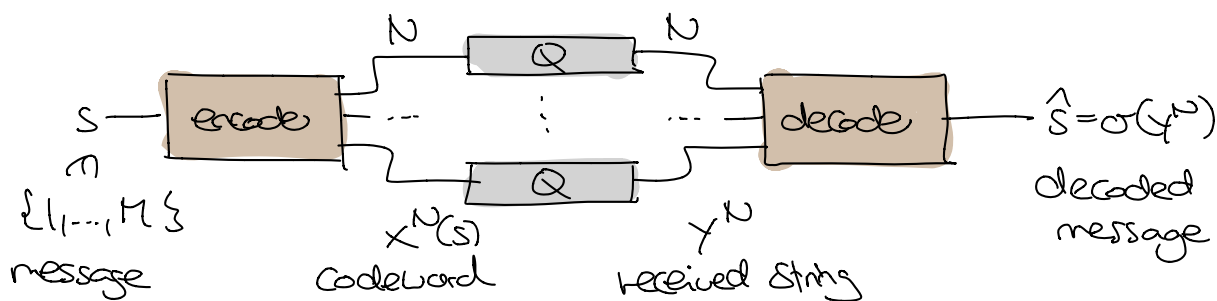
Recall: Capacity of channel $\mathcal{Q}(Y|X)$:

$$C(\mathcal{Q}) = \max_{P(x)} I(X; Y) \leftarrow \text{computed for } P(x, y) = P(x)\mathcal{Q}(y|x)$$

e.g. $C = 1 - H(\{f, 1-f\})$ for binary symmetric channel



The noisy coding theorem states: The capacity is the "optimal" rate at which we can communicate "reliably" using \mathcal{Q} . Let's state this more precisely:



(N, K) -block code: $x^N: \{1, 2, \dots, M\} \rightarrow \mathcal{X}^N$ where $M = 2^K$

Decodes: $\sigma: \mathcal{X}^N \rightarrow \{1, 2, \dots, M\}$

Convenient to include failure (but can also just decode incorrectly)

→ distribution of decoded message when sending s :

$$P(\hat{S}|s) = \Pr(\hat{S} = \hat{s} | S = s) = \sum_{\substack{y^N \text{ s.t.} \\ \sigma(y^N) = \hat{s}}} \mathcal{Q}(y_1 | x_1(s)) \dots \mathcal{Q}(y_n | x_n(s))$$

Components of $x^N(s)$

Figures of merit:

* rate: $R := \frac{K}{N}$ bits per channel use

* average prob. of (block) error for uniform $S' \in \{1, \dots, M\}$:

$$P_B = \Pr(\hat{S} \neq S') = \frac{1}{M} \sum_{s=1}^M \sum_{\hat{s} \neq s} P(\hat{S}|s)$$

Similarly for general $P(S)$

* maximal probability of (block) error:

$$P_{BM} = \max_S \Pr(\hat{S} \neq S' | S' = s) = \max_S \sum_{\hat{s} \neq s} P(\hat{S}|s)$$

How are these related?

* Clearly: $P_{BM} \geq P_B$

* Conversely: Define $(N, k-1)$ -code by removing the $\frac{M}{2} = 2^{k-1}$ codewords with largest $\Pr(\hat{S} \neq S | S=s)$. "expurgation"

$\Rightarrow P_{BM}^{new} \leq 2 P_B$ & $R^{new} = R - \frac{1}{N}$

Pf: Otherwise, original code had $> \frac{M}{2}$ codewords with $\Pr(\hat{S} \neq S | S=s) > 2P_B$
 $\Rightarrow P_B = \frac{1}{M} \sum_s \Pr(\hat{S} \neq S | S=s) > \frac{1}{2} \cdot 2P_B = P_B \quad \square$

enough to show for P_B instead of P_{BM}

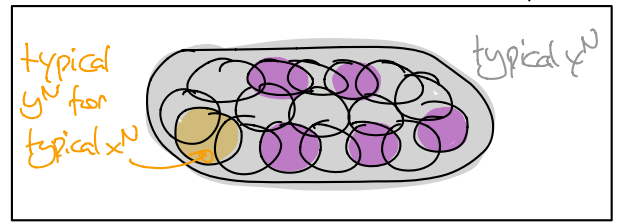
Shannon's noisy coding theorem: Let $Q(y|x)$ channel and $0 < \epsilon < 1$.

- (A) If $\tilde{R} < C(Q)$: $\exists N_0 \forall N \geq N_0$: $\exists (N, k)$ -code & decode with $\frac{k}{N} \geq \tilde{R}$ and $P_{err} \leq \epsilon$
- (B)? Thursday!

Intuition: Choose random codewords $X^N(s) \stackrel{i.i.d.}{\sim} P(x)$

typical channel outputs = ?

- in total $\sim 2^{NH(Y)}$
 - for typical codeword $\sim 2^{NH(Y|X)}$
- not so clear?



\hookrightarrow can choose $\sim \frac{2^{NH(Y)}}{2^{NH(Y|X)}} = 2^{NI(X;Y)}$ also not so clear? with little overlap

\hookrightarrow do this for $P(x)$ that achieves capacity of noisy typewriter!

Let's make this precise ...

Jointly typical set for $P(x,y)$:

$$J_{N,\epsilon}(P) = \left\{ (x^N, y^N) \text{ s.t. } x^N \in T_{N,\epsilon}(P_x), y^N \in T_{N,\epsilon}(P_y) \text{ and } (x^N, y^N) \in T_{N,\epsilon}(P_{xy}) \right\}$$

e.g. $|\frac{1}{N} \log \frac{1}{P(x^N, y^N)} - H(x,y)| \leq \epsilon$

Properties:

⑥ For all $(x^N, y^N) \in \mathcal{J}_{N, \epsilon}$: $2^{-N(H(X) + \epsilon)} \leq P(x^N) \leq 2^{-N(H(X) - \epsilon)}$
 (by definition) $2^{-N(H(XY) + \epsilon)} \leq P(x^N, y^N) \leq 2^{-N(H(XY) - \epsilon)}$

⑦ $\#\mathcal{J}_{N, \epsilon} \leq 2^{N(H(XY) + \epsilon)}$ (even holds for $T_{N, \epsilon}(R_{XY})$)

⑧ If $(X^N, Y^N) \stackrel{i.i.d.}{\sim} P(x, y)$: $\Pr((X^N, Y^N) \in \mathcal{J}_{N, \epsilon}) \rightarrow 1$ as $N \rightarrow \infty$ ← X_i & Y_i correlated via $P(x, y)$

Pf: $\Pr((X^N, Y^N) \notin \mathcal{J}_{N, \epsilon}) = \Pr(X^N \notin T_{N, \epsilon}(R_X) \text{ OR } \dots \text{ OR } \dots)$
 $\leq \Pr(X^N \notin T_{N, \epsilon}(R_X)) + \dots + \dots$ and each term $\rightarrow 0$.

⑨ If $\tilde{X}^N \stackrel{i.i.d.}{\sim} P(x)$ & $\tilde{Y}^N \stackrel{i.i.d.}{\sim} P(y)$ independent: $\Pr((\tilde{X}^N, \tilde{Y}^N) \in \mathcal{J}_{N, \epsilon}) \leq 2^{-N(I(X:Y) - 3\epsilon)}$ ← \tilde{X}_i indep. from \tilde{Y}_i

Pf: $\text{LHS} \stackrel{\text{independence}}{=} \sum_{(x^N, y^N) \in \mathcal{J}_{N, \epsilon}} P(x^N) P(y^N) \stackrel{\text{⑥} + \text{⑦}}{\leq} \#\mathcal{J}_{N, \epsilon} \cdot 2^{-N(H(X) - \epsilon)} \cdot 2^{-N(H(Y) - \epsilon)}$
 $\leq 2^{-N(I(X:Y) - 3\epsilon)}$ □

On Thursday we will use this to prove the noisy coding theorem!