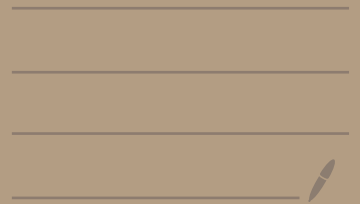
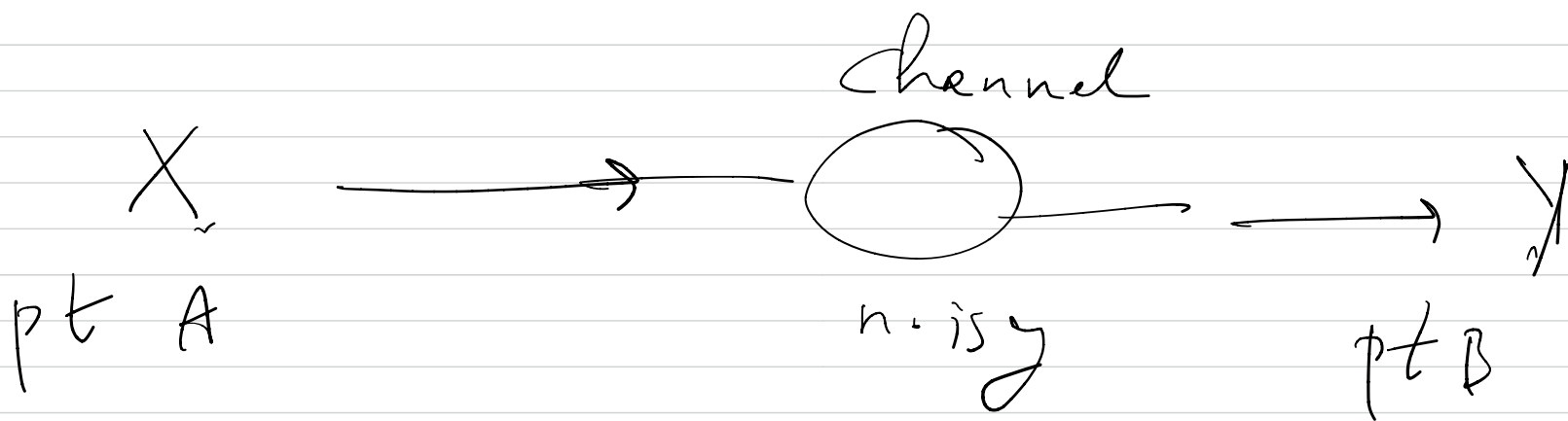


Information Theory &

Coding

Sept. 14th 2020





$$p(y|x) = P(Y=y | X=x)$$

↑

- probability kernel
- transition probability.

Source Coding / Data Compression

alphabet \mathcal{U} , $u \in \mathcal{U}$, $|\mathcal{U}| < \infty$
letter

U is a random variable taking values in \mathcal{U} .

a source code is a mapping

$$c: \mathcal{U} \rightarrow \{0, 1\}^* = \{\text{null}, 0, 1, 00, 01, 10, 11, 000, 001, \dots\}$$

= set of all finite (binary) sequences.

$$c_n: \mathcal{U}^n \rightarrow \{0, 1\}^*$$

ex: $\mathcal{U} = \{a, b, c, d\}$

$$c(a) = 0$$

$$c(b) = 0$$

$$c(c) = 1$$

$$c(d) = 0$$

Def: a code is called injective (non-singular)

if $u \neq v \Rightarrow c(u) \neq c(v)$

singular = non-injective

= bad.

= stupid

Ex: $\mathcal{U} = \{a, b, c, d\}$

$$\begin{array}{l} c(a) = 0 \\ c(b) = 00 \\ \hline c(c) = 1 \\ c(d) = 10 \end{array}$$

$$c^*(aa) = c^*(b)$$

not so.

$$c(\underline{d}) = c(\underline{a})c(\underline{a})$$

Def: Given $c: \mathcal{U} \rightarrow \{0,1\}^*$ let us define

$$c^*: \mathcal{U}^* \rightarrow \{0,1\}^* \quad a)$$

$$c^*(u_1 \dots u_n) = \underbrace{c(u_1) c(u_2) c(u_3) \dots c(u_n)}_{\text{Concatenation}}$$

$$c^n: \mathcal{U}^n \rightarrow \{0,1\}^* \quad a)$$

$$c^n(u_1 \dots u_n) =$$

Def. c is called uniquely-decodable

if c^* is injective.

Def. given a code $c: \mathcal{U} \rightarrow \{0,1\}^*$, define

$$\underline{\text{KraftSum}(c)} = \sum_{u \in \mathcal{U}} 2^{-\text{length}(c(u))}$$

for the code in the last example $KS = \frac{3}{2}$.

Thm: if $c: \mathcal{U} \rightarrow \{0,1\}^*$ is injective then

$$\underline{\text{KraftSum}(c) \leq \log_2(1 + |\mathcal{U}|)}$$

Pf: Without loss of generality we can

assume that if $k = \text{length}(c(u))$, then

for all binary sequences b with $\text{length}(b) < k$

there is a letter v s.t. $c(v) = b$.

(otherwise if b is not a codeword

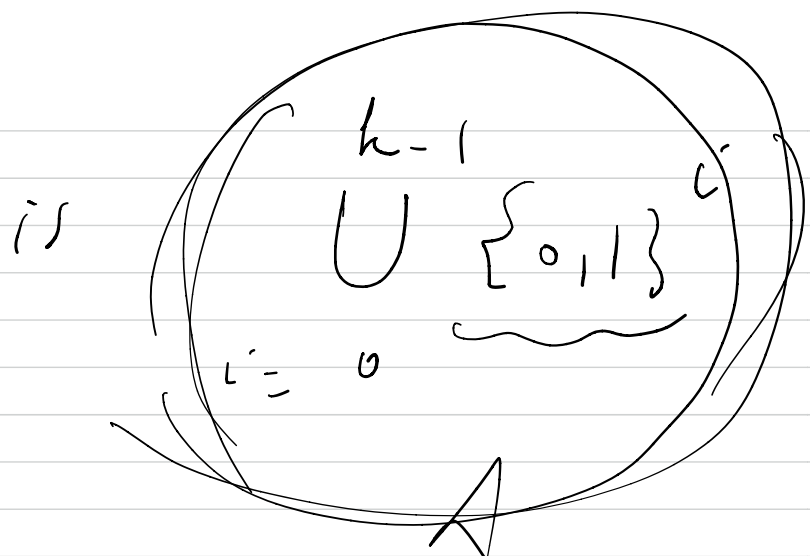
then we can set $c(u) = b$, without

violating injectivity, and this modification

↑ KS .) Consequently we can assume

that, with $k = \text{length of the longest}$

codeword, the set of codewords

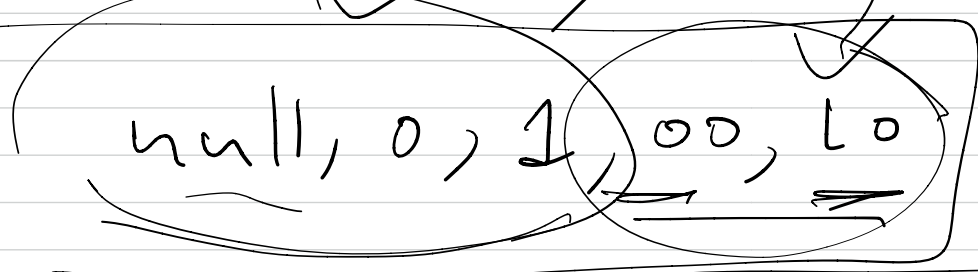


\cup (subset of $\{0,1\}^k$)

cardinality = r
 $1 \leq r \leq 2^k$

~~0, 00, 10, 11, 0000~~

→ 0, 00, 10, 11, (null) → 0, 1, 1



$$|U| = 1 + 2 + \dots + 2^{k-1} + r$$

$$= 2^k - 1 + r$$

$$K.S = \sum_{i=0}^{k-1} 2^i \cdot 2^{-i} + r 2^{-k}$$

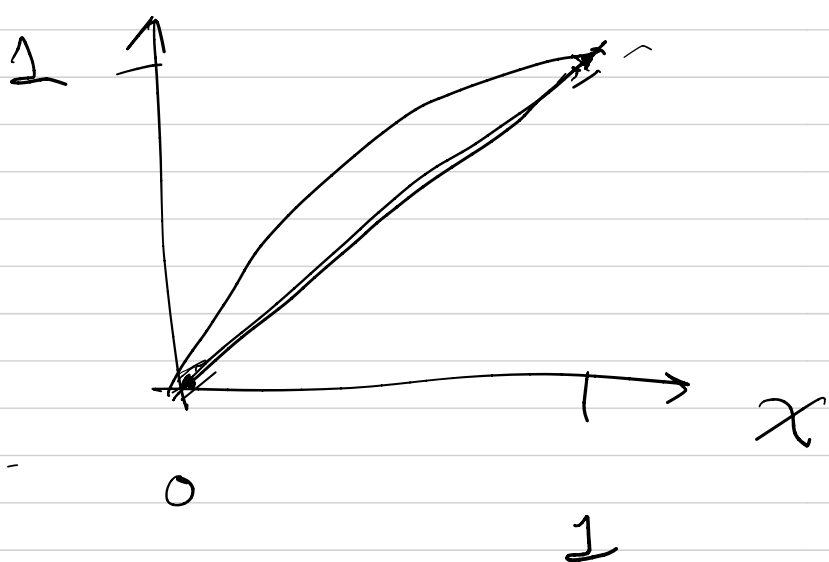
$$= k + r 2^{-k} = x$$

We are supposed to show $k + x \leq \log_2(|U|)$

$$\begin{aligned}
 k + \underbrace{x}_{\leftarrow} &\leq \log_2(2^k - 1 + 1 + r) \\
 &= \log_2(2^k + r) \\
 &= \log_2(2^k (1 + \underbrace{r 2^{-k}}_x))
 \end{aligned}$$

$$= k + \underbrace{\log_2(1+x)}_{\leftarrow}$$

$$0 < x \leq 1$$



$$x \rightarrow \log_2(1+x) - x$$

is concave

$$\Delta = 0 \text{ at } x=0, 1$$

\Rightarrow must be ≥ 0 in

between. //

Suppose now U is a RV taking values in \mathcal{U} , & we have a code $c: \mathcal{U} \rightarrow \{0,1\}^*$

$$E[\text{length } c(U)]$$

$$= \sum_u p(u) \text{length}(c(u))$$

$$= \sum_u p(u) \log_2 \frac{p(u)}{2^{-\text{length}(c(u))}}$$

$$= \sum_u p(u) \log \frac{1}{p(u)} - \sum_u p(u) \log_2 \frac{2^{-\text{length}(c(u))}}{p(u)}$$

$$\geq \sum_u p(u) \log \frac{1}{p(u)} - \log_2 \sum_u p(u) \frac{2^{-\text{length}(c(u))}}{p(u)}$$

concavity of \log

$$= \sum_u p(u) \log \frac{1}{p(u)} \Rightarrow \log_2(\text{Kraftsum}(c))$$

intrinsic to \mathcal{U}

Summary:

$$E(\text{length } c(U)) \geq H(U) - \underbrace{\log_2(K_{\text{raftSum}}(c))}_{\text{entropy of the source}}$$

$$\stackrel{\Delta}{=} \sum_u p(u) \log_2 \frac{1}{p(u)}$$

Then: if c is injective then,

$$E(\text{length } c(U)) \geq H(U) - \underbrace{\log_2 \log_2(1 + |U|)}_{\text{entropy of the source}}$$

Then: if c is uniquely decodable then

$$E(\text{length}(c(U))) \geq H(U).$$

Pf: if c is uniquely decodable then

$c^n: U^n \rightarrow \{0,1\}^*$ defined as

$c^n(u_1 \dots u_n) = c(u_1) \dots c(u_n)$ is injective.

$$\underbrace{E(\overset{\text{length}}{c^n}(u_1 \dots u_n))}_{\text{length}} \geq \underbrace{H(u_1 \dots u_n)}_{\text{entropy of the source}} - \log_2 \log_2(1 + |U|^n)$$

$$\text{note: length } c^n(u_1 \dots u_n)$$

$$= \text{length } c(u_1) + \text{length } c(u_2) + \dots + \text{length } c(u_n)$$

$$E(\text{length } c^n(u_1 \dots u_n))$$

$$= E(\text{length } c(u_1)) + \dots + E(\text{length } c(u_n))$$

$$= n E(\text{length } c(u))$$

$$H(u_1 \dots u_n) = \sum_{u_1 \dots u_n} p(u_1 \dots u_n) \log_2 \frac{1}{p(u_1 \dots u_n)}$$

$$p(u_1 \dots u_n) = p(u_1) p(u_2) \dots p(u_n) \quad \text{so}$$

$$\log_2 \frac{1}{p(u_1 \dots u_n)} = \log_2 \frac{1}{p(u_1)} + \log_2 \frac{1}{p(u_2)} + \dots + \log_2 \frac{1}{p(u_n)}$$

$$H(u_1 \dots u_n) = \sum_{u_1, u_2, \dots, u_n} p(u_1) p(u_2) \dots p(u_n) \log_2 \frac{1}{p(u_1)} + \dots$$

$$= \sum_{u_1} p(u_1) \log_2 \frac{1}{p(u_1)} + \dots$$

$$= H(u) + H(u) + \dots + H(u) = n H(u)$$

So we have

$$n E(\text{length } c(u)) \geq n H(u) - \log_2 \log_2 (1 + |U|^n)$$

$$\Rightarrow \underbrace{E(\text{length } c(u))}_{\text{}} \geq \underbrace{H(u)}_{\text{}} - \underbrace{\frac{1}{n} \log_2 \log_2 (1 + |U|^n)}_{\text{}} \quad \left(\frac{1}{n} \log_2 \log_2 (1 + |U|^n) \right)$$

By taking limit as $n \rightarrow \infty$ we get $O\left(\frac{1}{n} \log_2 n\right)$

$$\underline{E(\text{length } c(u))} \geq \underline{H(u)} \quad //$$