

## Prefix-free code (instantaneous code)

code is prefix-free if no codeword is a prefix of any other

binary code tree: all codewords must be leaves for prefix-freeness.

PF code is full if no new codewords can be added without destroying PF property.

How do we know if we can construct a PF code for a given source with a particular codeword length?

## Kraft Inequality (1949)

Let  $X = \{x_1, \dots, x_M\}$  and let  $l(x_1), \dots, l(x_M)$  be codeword lengths.

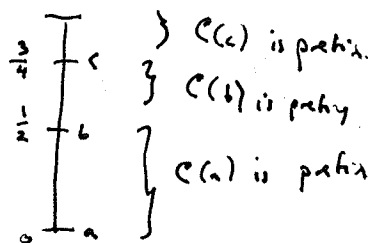
Then  $\sum_{j=1}^M 2^{-l(x_j)} \leq 1$  if and only if PF code exists.

PF code is full when holds with equality.

proof: ① model the space of codewords as unit interval  $[0, 1)$ , since any codeword can be represented as a binary expansion.

$$\begin{aligned} 0.b_1 b_2 \dots &= b_1 2^{-1} + b_2 2^{-2} + \dots \\ &= \frac{1}{2} b_1 + \frac{1}{4} b_2 + \dots \end{aligned}$$

$$\begin{aligned} a &\rightarrow 0 & 0 \\ b &\rightarrow 10 & \frac{1}{2} \\ c &\rightarrow 11 & \frac{3}{4} \end{aligned}$$



② so a given codeword  $c(x_j) = c_j$  is prefix for all codewords in an interval of size  $2^{-l_j}$  starting at  $c_j$ .

③ prefix-free property  $\Leftrightarrow$  intervals don't overlap.

$\rightarrow$  check that PF satisfies  $\sum_{j=1}^M 2^{-l_j} \leq 1$

Can obtain PF code by reading off interval boundaries.

optimizing PF codes

$$\text{minimize } \sum_{j=1}^M p_j l_j \quad \text{s.t. } \sum_{j=1}^M 2^{-l_j} \leq 1 \quad \text{our choice } l_1, \dots, l_M.$$

to get lower bound, ignore integer constraints (so Kraft inequality holds with equality)

using Lagrange multipliers, get

$$l_j = -\log p_j$$

$$\bar{L} \approx \sum_j p_j l_j = -\sum_j p_j \log p_j = H(X).$$

to get upper bound, for original integer program, round real-valued optimal lengths up.

$$l_j = \lceil -\log p_j \rceil \quad \text{which is Shannon-Fano-Elias code.} \\ < 1 - \log p_j$$

$$\bar{L} = \sum_j p_j l_j < \sum_j p_j (1 - \log p_j) = 1 - \sum_j p_j \log p_j = H(X) + 1$$

$$\text{so } H(X) \leq \bar{L} < H(X) + 1$$

Asymptotic Equipartition Property (AEP), a restatement of the law of large numbers

Let  $-\frac{1}{n} \log p(\vec{x}) = -\frac{1}{n} \sum_{k=1}^n \log p(x_k)$  be the empirical entropy of a particular source sequence. Thought to be generated iid according to r.v.  $X$  with entropy  $H(X)$ .

The typical set is defined as set of sequences whose empirical entropy is close to the true entropy.  $A_\epsilon^{(n)}$  is  $\vec{x} \in \mathcal{X}^n$  s.t.

$$\left| -\frac{1}{n} \log p(\vec{x}) - H(X) \right| \leq \epsilon \quad \text{or}$$

$$2^{-n(H(X)+\epsilon)} \leq p(\vec{x}) \leq 2^{-n(H(X)-\epsilon)}$$

AEP: ① if  $\vec{x} \in A_\epsilon^{(n)}$  then  $2^{-n(H(X)+\epsilon)} \leq p(\vec{x}) \leq 2^{-n(H(X)-\epsilon)}$

② for sufficiently large  $n$ ,  $\Pr[A_\epsilon^{(n)}] > 1-\epsilon$

③ for sufficiently large  $n$ ,

$$(1-\epsilon) 2^{n(H(X)-\epsilon)} \leq |A_\epsilon^{(n)}| \leq 2^{n(H(X)+\epsilon)}$$

Total number of typical sequences  $\approx 2^{nH(X)}$  which is generally much less than all possible sequences, if  $H(X) < \log|\mathcal{X}|$ .

when  $n$  large, can essentially think of sequence  $X$  as being drawn uniformly from typical set

### Back to block coding

Let  $X_1, X_2, \dots, X_n$  be iid r.v. from  $p(x)$ .

want to find short descriptions for such sequences.

#### Achievable scheme:

① divide sequences in  $X^n$  into  $A_\epsilon^{(n)}$  and its complement, order two subsets in some fixed way.

② represent typical sequences by indexing with binary codewords of length no more than  $n(H+\epsilon)+1$ , since less than  $2^{n(H+\epsilon)}$  elements in  $A_\epsilon^{(n)}$

→ prefix with a 0, so total length  $\leq n(H+\epsilon)+2$ .

③ represent each sequence not in  $A_\epsilon^{(n)}$  using no more than  $n \log |X| + 1$  bits, and prefix with 1.

#### Performance of achievable scheme

let  $l(x^n)$  be length of codeword for  $x^n$ , suppose  $n$  sufficiently large for  $Pr[A_\epsilon^{(n)}] \geq 1-\epsilon$ .

$$\begin{aligned}
E[l(x^n)] &= \sum_{x^n} p(x^n) l(x^n) = \sum_{x^n \in A_\epsilon^{(n)}} p(x^n) l(x^n) + \sum_{x^n \notin A_\epsilon^{(n)}} p(x^n) l(x^n) \\
&\leq \sum_{x^n \in A_\epsilon^{(n)}} p(x^n) (n(H+\epsilon)+2) + \sum_{x^n \notin A_\epsilon^{(n)}} p(x^n) (n \log |X| + 2) \\
&= Pr[A_\epsilon^{(n)}] (n(H+\epsilon)+2) + (1-Pr[A_\epsilon^{(n)}]) (n \log |X| + 2) \\
&\leq n(H+\epsilon) + \epsilon n (\log |X|) + 2
\end{aligned}$$

$$\text{letting } \epsilon' = \epsilon + \epsilon \log |X| + \frac{2}{n}$$

$$= n(H + \epsilon')$$

So achievable scheme has  $E\left[\frac{1}{n} \ell(X^n)\right] = H(X) + \epsilon$  for large  $n$ .

Also converse: if we use block code with blocklength  $n$  and coding rate less than  $H(X) - \zeta$  where  $\zeta > 0$  (doesn't change with  $n$ ) then error probability goes to 1 as  $n \rightarrow \infty$ .

proof idea: there isn't a smaller set than the typical set that contains most of the probability mass, measured in exponential sense.  $(a_n \doteq b_n \text{ means } \lim_{n \rightarrow \infty} \frac{1}{n} \log \frac{a_n}{b_n} = 0)$