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Information Theory

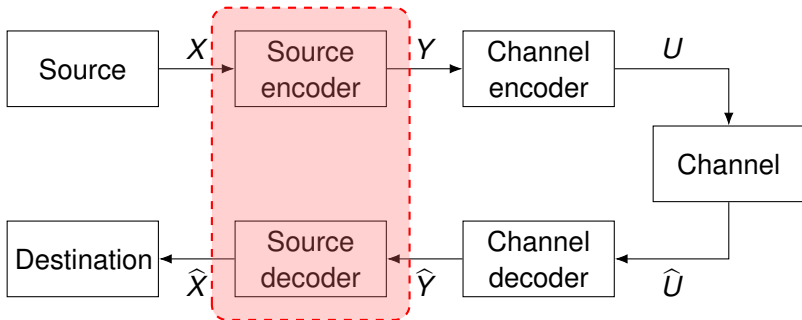
Lecture 3

Optimal source coding and Kraft inequality

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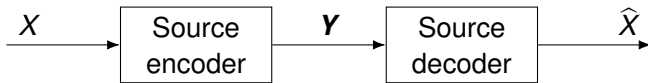


Communication system



Shannon's model for a communication system

Source coding



Definition

A **source code** is a mapping from the random variable $X \in \{x_1, \dots, x_k\}$ to a vector of variable length ℓ , $\mathbf{y} = (y_1 y_2 \dots y_\ell)$, where $y_i \in \mathbb{Z}_D$ are drawn from a D -ary alphabet.

The length of the codeword corresponding to x is denote ℓ_x .

The efficiency of a code is shown by the average codeword length,

$$L = E[\ell_x] = \sum_x p(x) \ell_x$$

Code classification

Definition

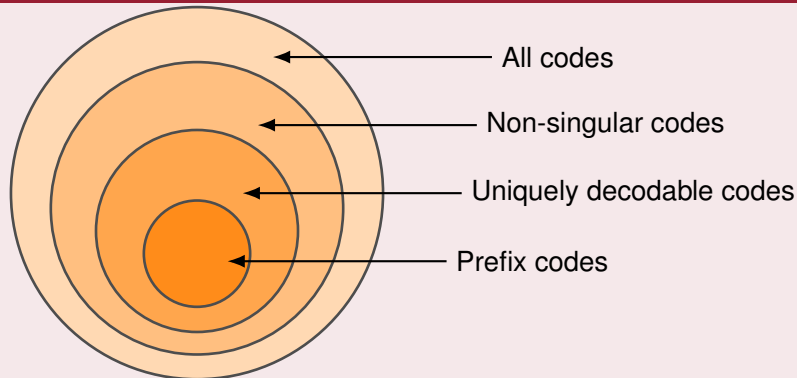
A code is said to be

- **Non-singular** if each source symbol is mapped to a distinct code vector.
- **Uniquely decodable** if each sequence of source symbols is mapped to a unique sequence of code symbols, i.e. the mapping from source sequences is non-singular.
- **Prefix** (or **instantaneous**) if no codeword is a prefix of another codeword.



Code classification

Illustration of code classification



Prefix codes

D -ary tree

- A D -ary tree is a tree where each node has either D or 0 branches.
- A **full** D -ary tree of depth n is a tree with depth n and D^n end nodes (leaves).

Code representation

A prefix code can be represented in a D -ary tree, with branches representing symbols. The depth of a path is the length of the codeword.

Optimal code

Definition

An optimal prefix code is a prefix code that minimizes the expected codeword length

$$L = E[\ell_x] = \sum_x p(x) \ell_x$$

over all prefix codes for the same source alphabet.

Lemma (Path length lemma)

In a tree representation of a prefix code, the average codeword length $L = E[\ell_x]$ equals the sum of probabilities for the inner nodes, including the root.

Kraft inequality

Theorem (Kraft inequality)

There exists a *D-ary prefix code* with codeword lengths $\ell_1, \ell_2, \dots, \ell_k$ if and only if

$$\sum_{i=1}^k D^{-\ell_i} \leq 1$$

Kraft inequality — Sketch of proof

Prefix code \Rightarrow Inequality

- Assume prefix code with $\max_x \ell_x = \ell_{\max}$
- Represent the codewords in a full tree with $D^{\ell_{\max}}$ leaves
- Remove subtree after codewords

$$\Rightarrow \sum_x D^{\ell_x} \leq 1$$

Kraft inequality — Sketch of proof

Inequality \Rightarrow Prefix code

- Assume $\sum_i D^{\ell_{x_i}} \leq 1$ and $\ell_{x_1} \leq \ell_{x_2} \leq \dots \leq \ell_{x_k} = \ell_{\max}$
- Start with full tree of depth $D^{\ell_{\max}}$
- Put in shortest codewords and remove subtree after
- After $j < k$ steps there are unused nodes left at depth $D^{\ell_{\max}}$

At $j = k - 1$ there are at least one node left

\Rightarrow The code can be constructed

Kraft inequality

Theorem (McMillan inequality)

There exists a D -ary uniquely decodable code with codeword lengths l_1, l_2, \dots, l_k if and only if

$$\sum_{i=1}^k D^{-l_i} \leq 1$$

Uniquely decodable codes will never be better than prefix codes.

Bounds on L

Theorem

The expected codeword length $L = E[\ell_x]$ of a prefix code is lower bounded by the entropy of the source, i.e.

$$L \geq H_D(X) = \frac{H(X)}{\log D}$$

with equality if and only if the optimal codeword lengths $\ell_x = -\log_D p(x)$ is used.

Theorem

For every random variable X there exists a D -ary prefix code such that

$$L < H_D(X) + 1$$

Optimal codeword length

Theorem

Given a random variable X , the average codeword length for an optimal D -ary prefix code satisfies

$$H_D(X) \leq L < H_D(X) + 1$$

with equality to the left if and only if $\ell_x = \log_D p(x)$.