# **Information and Communication Theory**

Lecture 3

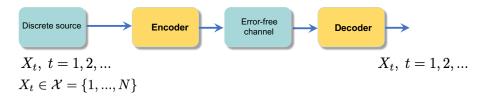
# **Optimal Coding**

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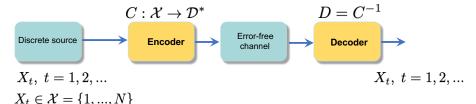
2023

### Source Coding



- Lossless encoding: output of the decoder equal to that of the source.
- Assumption: when encoding  $X_t$ , its distribution is known:
  - ✓ For memoryless sources, this is just  $f_X$ ;
  - $\checkmark$  For Markov sources, this is  $f_{X_t|X_{t-1},...}$
- Without loss of generality, we simply write  $f_X$ .
- Goal: economy, that is, use the channel as little as possible.

## Variable-Length Coding

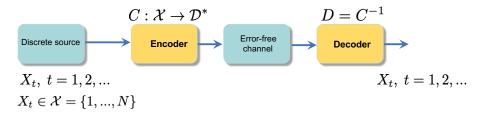


- Code uses D-ary alphabet  $\mathcal{D} = \{0, 1, ..., D-1\}.$
- Typically, binary coding, D = 2,  $\mathcal{D} = \{0, 1\}$ .
- Variable-length encoding:  $\mathcal{D}^*$  is the Kleene closure of  $\mathcal{D}$ :

$$\mathcal{D}^* = \{\text{all finite strings of symbols of } \mathcal{D}\} = \bigcup_{n=0}^{\infty} \mathcal{D}^n$$

 $\begin{array}{l} \bullet \ \ \mathsf{Example:} \ \ \mathsf{for} \ \mathcal{D} = \{0,1\}, \\ \\ \mathcal{D}^* = \{\epsilon,0,1,00,01,10,11,000,001,010,011,100,101,110,111,0000,\ldots\} \end{array}$ 

### Non-Singular and Uniquely Decodable Codes



ullet For  $C^{-1}$  to exist: non-singular code (C injective). For any  $x,y\in\mathcal{X}$ ,

$$x \neq y \Rightarrow C(x) \neq C(y)$$

• To be useful for a sequence of symbols, this is not good enough.

Example: 
$$\{C(1)=0,\ C(2)=10,\ C(3)=01\}$$
 is non-singular Received sequence:  $010$ ; is it  $C(1)C(2)$  or  $C(3)C(1)$ ? Impossible to know!

Codes that do not have this problem are called uniquely decodable.

#### Instantaneous Codes

• Consider a uniquely decodable code:

$${C(1) = 01, C(2) = 11, C(3) = 00, C(4) = 110}$$

• How to decode the sequence  $11\underline{00...0011?}$ 

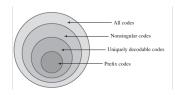
✓ If 
$$n$$
 is even:  $C^{-1}(11\underbrace{00....00}_{n \, \text{zeros}}11) = 2\underbrace{3....3}_{n/2}2$ 

$$\checkmark \ \ \text{If} \ n \ \text{is odd:} \ C^{-1}(11\underbrace{00....00}_{n \ \text{zeros}}11) = 4\underbrace{3....3}_{\underbrace{n-1}{2}}2$$

- To decode the first symbol, we many need to wait for many others.
- A code that does not have this problem is called instantaneous.

#### Instantaneous Codes

If no codeword is prefix of another, decoding is instantaneous.
 Other names: prefix codes, prefix-free codes.



• Length function:

$$l_C(x) = \operatorname{length}(C(x))$$

Expected length

$$L(C) = \mathbb{E}[l_C(X)] = \sum_{x \in \mathcal{X}} f_X(x) \, l_C(x)$$

#### Example:

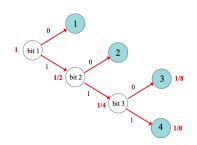
х	f <sub>X</sub> (x)	C(x)	$l_{\rm C}({\rm x})$	
1	1/2	0	1	
2	1/4	10	2	
3	1/8	110	3	
4	1/8	111	3	

$$L(C) = 7/4$$
 bits/symbol

### Instantaneous Codes: Tree Representation

- Instantaneous code: no codeword is prefix of another.
- Decoding instantaneous code: path from root to leaf of a tree:

х	f <sub>X</sub> (x)	C(x)	$l_{\rm C}({\rm x})$
1	1/2	0	1
2	1/4	10	2
3	1/8	110	3
4	1/8	111	3



- For D-ary codes: D-ary trees.
- ullet L(C) is the sum of the probabilities of the inner nodes. (show why)

### Instantaneous Codes: Kraft-McMillan Inequality

• If C is a D-ary instantaneous code, it necessarily satisfies

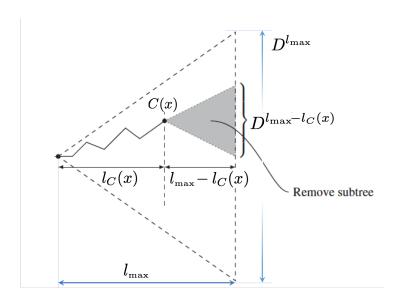
$$\sum_{x \in \mathcal{X}} D^{-l_C(x)} \le 1 \qquad \text{(KMI)}$$

- ...i.e., if some words are short others have to be long!
- Proof:
  - ✓ let  $l_{\text{max}} = \max\{l_C(1), ..., l_C(N)\}$  (length of the longest word).
  - ✓ there are  $D^{l_{\text{max}}}$  words of length  $l_{\text{max}}$ .
  - ✓ for each word C(x), there are  $D^{l_{\max}-l_C(x)}$  words of length  $l_{\max}$  that have C(x) as prefix;
  - $\checkmark$  the sets of length- $l_{
    m max}$  words that have each word as prefix are disjoint.

$$\sum_{x \in \mathcal{X}} D^{l_{\max} - l_C(x)} \leq D^{l_{\max}} \overset{\text{divide}}{\longrightarrow} \overset{\text{by } D^{l_{\max}}}{\longrightarrow} \sum_{x \in \mathcal{X}} D^{-l_C(x)} \leq 1$$

• Important: the KMI is a necessary, not sufficient, condition. (why?)

# Kraft-McMillan Inequality: Graphical Proof



# Source Coding Theorem

- Source  $X \in \mathcal{X} = \{1, ..., N\}$  with probability mass function  $f_X$ .
- For any collection of N positive integers,  $l_1,...,l_N$ ,

$$(\mathsf{KMI}) \ \sum_{x \in \mathcal{X}}^{N} D^{-l_x} \le 1 \ \Rightarrow \ \sum_{x \in \mathcal{X}} f_X(x) \, l_x \ge H(X).$$

 $\bullet \ \, \text{Proof: let } q(x) = \frac{D^{-l_x}}{A} > 0 \text{, where } A = \sum_{x \in \mathcal{X}} D^{-l_x} \leq 1; \, \sum_{x \in \mathcal{X}} q(x) = 1$ 

$$0 \le D_{\mathsf{KL}}(f_X \parallel q) = \sum_{x \in \mathcal{X}} f_X(x) \log_D \frac{f_X(x)}{q(x)}$$
$$= \underbrace{\sum_{x \in \mathcal{X}} f_X(x) \log_D f_X(x)}_{-H_D(X)} + \underbrace{\log_D A}_{x \in \mathcal{X}} \sum_{x \in \mathcal{X}} f_X(x) + \sum_{x \in \mathcal{X}} f_X(x) l_x$$

...equality iff A=1 and  $q(x)=f_X(x)\Leftrightarrow l_x=-\log_D f_X(x)$  (only possible if integers).

### Source Coding Theorem

- Source  $X \in \mathcal{X} = \{1, ..., N\}$  with probability mass function  $f_X$ .
- Corollary of the result in previous slide:

$$C$$
 is instantaneous  $\Rightarrow$  KMI  $\Rightarrow$   $\sum_{x \in \mathcal{X}} f_X(x) \, l_C(x) \geq H_D(X)$  expected code-length  $L(C)$ 

- ...with equality if and only if  $l_C(x) = -\log f_X(x)$ .
- Equality is only possible if  $-\log_D f_X(x)$  are integers.
- ullet Shannon-Fano code (SFC): just take  $l_C^{\rm SF}(x) = \lceil -\log_D f_X(x) 
  ceil$
- Clearly, the SFC satisfies the KMI  $(\lceil u \rceil \geq u$ , for any  $u \in \mathbb{R})$

$$\sum_{x \in \mathcal{X}} D^{-l_C^{\mathsf{SF}}(x)} \le \sum_{x \in \mathcal{X}} D^{\log_D f_X(x)} = \sum_{x \in \mathcal{X}} f_X(x) = 1$$

### **Optimal Code**

- Source  $X \in \mathcal{X} = \{1, ..., N\}$  with probability mass function  $f_X$ .
- Optimal code lengths:

$$\begin{split} \left(l_1^{\mathsf{optimal}},...,l_N^{\mathsf{optimal}}\right) &= \arg\min_{(l_1,...,l_N)} \sum_{x \in \mathcal{X}} f_X(x) \; l_x \\ & \mathsf{subject to} \;\; l_1,...,l_N \in \mathbb{N} \\ & \sum_{x \in \mathcal{X}} D^{-l_x} \leq 1 \end{split}$$

ullet Optimal code:  $C^{\text{optimal}}$  is any instantaneous code with

$$l_{C^{\mathsf{optimal}}}(x) = l_x^{\mathsf{optimal}}, \text{ for } x \in \mathcal{X}$$

- Because it satisfies the KMI,  $L(C^{\text{optimal}}) \ge H(X)$ .
- Because it is optimal,  $L(C^{\text{optimal}}) \leq L(C^{\text{SF}})$

# Bounds on Optimal Code-length

- $\bullet \ \, \text{Because it is optimal, } L(C^{\text{optimal}}) \leq L(C^{\text{SF}})$
- Because  $\lceil u \rceil < u+1$ , for any  $u \in \mathbb{R}$ ,

$$L(C^{\text{optimal}}) \le L(C^{\text{SF}}) = \sum_{x \in \mathcal{X}} f_X(x) \lceil -\log_D f_X(x) \rceil$$

$$< \sum_{x \in \mathcal{X}} f_X(x) (-\log_D f_X(x) + 1) = H(X) + 1$$

- In summary:  $H(X) \le L(C^{\text{optimal}}) < H(X) + 1$
- Code efficiency:  $\rho_C = \frac{H(X)}{L(C)}$ .
- Ideal code:  $\rho_C=1$ . Important: ideal  $\stackrel{\Rightarrow}{\not=}$  optimal

## Coding With a Wrong Distribution

- Source  $X \in \mathcal{X} = \{1, ..., N\}$  with probability mass function  $f_X$ .
- Build Shannon-Fano code assuming  $g_X$ :  $l_C(x) = \lceil -\log g_X(x) \rceil$
- Lower bound:

$$L(C) = \sum_{x \in \mathcal{X}} \lceil -\log g_X(x) \rceil f_X(x)$$

$$\geq -\sum_{x \in \mathcal{X}} f_X(x) \log g_X(x)$$

$$= \sum_{x \in \mathcal{X}} f_X(x) \log \frac{f_X(x)}{g_X(x) f_X(x)}$$

$$= -\sum_{x \in \mathcal{X}} f_X(x) \log f_X(x) + \sum_{x \in \mathcal{X}} f_X(x) \log \frac{f_X(x)}{g_X(x)}$$

$$\xrightarrow{H(X)} \xrightarrow{D_{KL}(f_X || g_X)}$$

# Coding With a Wrong Distribution

- Source  $X \in \mathcal{X} = \{1, ..., N\}$  with probability mass function  $f_X$ .
- Build Shannon-Fano code assuming  $g_X$ :  $l_C(x) = \lceil -\log g_X(x) \rceil$
- Upper bound:

$$L(C) = \sum_{x \in \mathcal{X}} \lceil -\log g_X(x) \rceil f_X(x)$$

$$< \sum_{x \in \mathcal{X}} f_X(x) (-\log g_X(x) + 1)$$

$$= H(X) + D_{\mathsf{KL}}(f_X \parallel g_X) + 1$$

ullet Summarizing: if C is built from  $g_X$  and the true distribution is  $f_X$ 

$$H(X) + D_{\mathsf{KL}}(f_X \parallel g_X) \le L(C) < H(X) + D_{\mathsf{KL}}(f_X \parallel g_X) + 1$$

### Approaching the Bound: Source Extension

- Discrete stationary source  $X_t \in \mathcal{X} = \{1,...,N\}$
- Extension: group n consecutive symbols:  $(X_1,...,X_n) \in \{1,...,N\}^n$ .
- ullet The optimal code for the extended symbols  $(X_1,...,X_n)$  satisfies

$$H(X_1, ..., X_n) \le \underbrace{L(C_n^{\mathsf{optimal}})}_{\mathsf{bits/}(n \; \mathsf{symbols})} < H(X_1, ..., X_n) + 1$$

• Memoryless source:  $H(X_1,...,X_n) = n H(X_1)$ , thus

$$L\left(C_n^{\mathsf{optimal}}\right) \leq n\,H(X_1) + 1 \quad \Rightarrow \quad \underbrace{\frac{L\left(C_n^{\mathsf{optimal}}\right)}{n}}_{\mathsf{bits/symbol}} < H(X_1) + \frac{1}{n}$$

...via extension, expected code-length can arbitrarily approach the entropy.

• Non-memoryless source:  $H(X_1,...,X_n) < nH(X_1)$  and the result is even stronger.

- Huffman (1952) algorithm to obtain optimal codes.
- ullet Builds a D-ary tree, starting from the leaves, which are the symbols.
- Algorithm (for D=2; the generalizing to D>2 requires some care).
  - **1 Input**: a list of symbol probabilities  $(p_1, ..., p_N)$ .
  - Output: a binary tree with each symbol as a leaf.
  - Assign each symbol to a leaf of the tree.
  - Find the 2 smallest probabilities:  $p_i$  and  $p_j$ .
  - **5** Create the parent node for nodes i and j with probability  $p_i + p_j$ .
  - **1** Remove  $p_i$  and  $p_j$  from the list and insert  $p_i + p_j$ .
  - If the list of symbols has more than 2 probabilities, go back to step 4.
- As seen before, a binary tree corresponds to an instantaneous code.

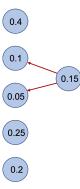


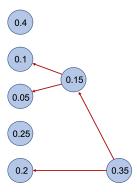


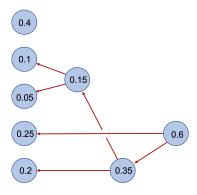


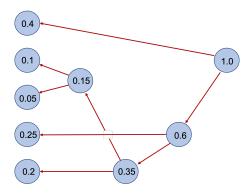




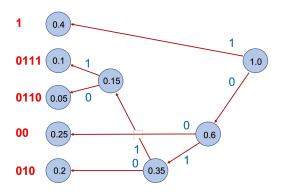




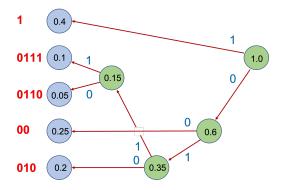




• Label the edges (arbitrarily) to obtain the code words



• Expected code-length: sum of the inner node probabilities:  $L(C)=1+0.6+0.35+0.15=2.1 \ {\rm bits/symbol}$ 



- Huffman codes are optimal; see proof in recommended reading.
- Converse is not true

Huffman code 
$$\stackrel{\Rightarrow}{\not=}$$
 optimal code

- In the case of ties, break them arbitrarily.
- For *D*-ary codes, merge *D* symbols to build a *D*-ary tree.
- For D-ary codes, optimality requires N=k(D-1)+1, where  $k\in\mathbb{N}$ . ...if not satisfied, just append zero-probability symbols.

#### Elias Codes for Natural Numbers

- Standard number representation is not uniquely decodable.
- Binary representation of natural numbers is not uniquely decodable. Example: C(3) = 11, C(21) = 10101, but decoding 1110101 is impossible; it could be C(14)C(5) or C(58)C(1).
- Length of binary representation for  $x \in \mathbb{N}$  is  $\lfloor \log_2 x \rfloor + 1$ . Example: C(13) = 1101 has length 4;  $\lfloor \log_2 13 \rfloor + 1 = \lfloor 3.70 \rfloor + 1 = 4$ .
- Elias coding:
  - √ instantaneous code for arbitrary natural numbers;
  - ✓ length not much worse than  $\lfloor \log_2 x \rfloor + 1$ .
- Useful not only for  $\mathbb N$ , but also for large alphabets  $\mathcal X=\{1,...,N\}$  with large and unknown N.

#### Elias Gamma Code

- Length of binary representation for  $x \in \mathbb{N}$  is  $\lfloor \log_2 x \rfloor + 1$ .
- Let  $C_2$  denote the standard binary representation.
- Elias gamma code:

x	$C_{\gamma}(x)$
1	1
2	010
4	00100
5	00101
7	00111
9	0001001
10	0001010
:	:
19	000010011
:	i i
147	000000010010011

$$C_{\gamma}(x) = \underbrace{0...0}_{|\log_2 x|_{\mathsf{zeros}}} C_2(x)$$

- Obviously instantaneous.
- Length:  $l_{C_{\gamma}}(x) = 2\lfloor \log_2 x \rfloor + 1.$
- Twice as long as  $C_2$ .

#### Elias Delta Code

- ullet Elias delta code:  $C_\delta(x) = C_\gamma(\lfloor \log_2 x \rfloor + 1) \ ilde{C}_2(x)$
- ullet  $ilde{C}_2$  is  $C_2$  without the leading 1 (e.g.  $C_2(9)=1001,\ ilde{C}_2(10)=001$ )
- Length:  $l_{C_\delta}(x) = l_{C_\gamma}(\lfloor \log_2 x \rfloor + 1) + \lfloor \log_2 x \rfloor$   $= 2 \lceil \log_2(\lceil \log_2 x \rceil + 1) \rceil + \lceil \log_2 x \rceil + 1$

x	$C_{\delta}(x)$
1	1
2	0100
3	0101
4	01100
7	01111
8	00100000
10	00100 <mark>010</mark>
:	:
19	00101 <mark>0011</mark>
:	i l
147	0001000 <mark>0010011</mark>

- Obviously instantaneous.
- For x > 32,  $l_{C_{\delta}}(x) < l_{C_{\gamma}}(x)$
- Approaches  $C_2$  for large x:

$$\lim_{x \to \infty} \frac{l_{C_{\delta}}(x)}{C_2(x)} = 1$$

## Recommended Reading

- T. Cover and J. Thomas, "Elements of Information Theory", John Wiley & Sons, 2006 (Chapter 5).
- M. Figueiredo, "Elias Coding for Arbitrary Natural Numbers", available at the course webpage in Fenix.