

INFORMATION THEORY & CODING

Week 5 : Source Coding 1

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October 18, 2022



Theorem (AEP)

“Almost all events are almost equally surprising.” Specifically, if X_1, X_2, \dots are i.i.d. $\sim p(x)$, then

$$-\frac{1}{n} \log p(X_1, X_2, \dots, X_n) \rightarrow H(X) \text{ in probability.}$$

Definition

The *typical set* $A_\epsilon^{(n)}$ is the set of sequences x_1, x_2, \dots, x_n satisfying

$$2^{-n(H(X)+\epsilon)} \leq p(x_1, x_2, \dots, x_n) \leq 2^{-n(H(X)-\epsilon)}.$$

Review Summary

Properties of the typical set

1. If $(x_1, x_2, \dots, x_n) \in A_\epsilon^{(n)}$, then
$$H(X) - \epsilon \leq -\frac{1}{n} \log p(x_1, x_2, \dots, x_n) \leq H(X) + \epsilon.$$
2. $\Pr[A_\epsilon^{(n)}] > 1 - \epsilon$ for n sufficiently large.
3. $|A_\epsilon^{(n)}| \leq 2^{n(H(X)+\epsilon)}$, where $|A|$ denotes the cardinality of the set A .
4. $|A_\epsilon^{(n)}| \geq (1 - \epsilon)2^{n(H(X)-\epsilon)}$ for n sufficiently large.

Theorem

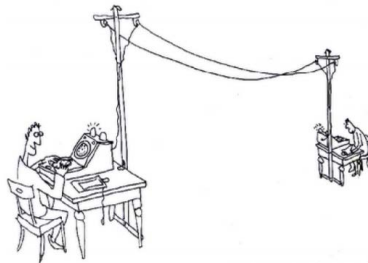
Let X^n be i.i.d. $\sim p(x)$. There exists a code that one-to-one maps sequences x^n of length n into binary strings with

$$E\left[\frac{1}{n} \ell(X^n)\right] \leq H(X) + \epsilon$$

for n sufficiently large.

Source Coding

Which horse won in the horse racing?



ABC TELEGRAPH EXHIBIT FOR THE NATIONAL ARCHIVES
INVENTIONS EXHIBITION TIM BUNKIN 7/10/8

Source Coding

source alphabet $X = \{0, 1, \dots, 7\}$

Which horse won in the horse racing?

code alphabet $D = \{0, 1\}$

alphabet extension

$D^* = \{0, 1, 00, \dots\}$

X	Pr	Code I	Code II
0	1/2	000	0
1	1/4	001	10
2	1/8	010	110
3	1/16	011	1110
4	1/64	100	111100
5	1/64	101	111101
6	1/64	110	111110
7	1/64	111	111111

$$H(X) = - \sum p_i \log p_i = 2\text{bits}$$

Which code is better?

Source Coding (Data Compression)

- We interpret that $H(X)$ is the best achievable data compression.
- We want to develop practical **lossless coding algorithms** that approach, or achieve the entropy limit $H(X)$.

Terminology

X	Pr	Code I	Code II
0	1/2	000	0
1	1/4	001	10
2	1/8	010	110
3	1/16	011	1110
4	1/64	100	111100
5	1/64	101	111101
6	1/64	110	111110
7	1/64	111	111111

- Source alphabet $\mathcal{X} = \{0, 1, 2, 3, 4, 5, 6, 7\}$.
- Code alphabet $\mathcal{D} = \{0, 1\}$.
- Codeword, e.g., 010 for $X = 2$ in Code 1.
- Codeword length, e.g., codeword length for Code 1 is 3.
- Codebook: all the codewords.

Notation (Alphabet Extension)

The set of all possible sequences based on a finite alphabet \mathcal{D} is denoted by \mathcal{D}^* . E.g.,

$$\mathcal{D} = \{0, 1\} \mapsto \mathcal{D}^* = \{0, 1, 00, 01, 10, 11, 000, \dots\}.$$

Definition (Source Code)

Let \mathcal{X} be the alphabet of a random variable X , and \mathcal{D} be the alphabet of code. A *source code* C for the random variable X is a map

$$\begin{aligned} C : \quad \mathcal{X} &\rightarrow \mathcal{D}^* \\ x &\mapsto C(x) \end{aligned}$$

where $C(x)$ is the codeword associated with x . Let $\ell(x)$ denote the length of $C(x)$.

Definition

The *expected length* $L(X)$ of a source code C for a random variable X with probability mass function $p(x)$ is

$$L(X) = E\ell(X) = \sum_{x \in \mathcal{X}} p(x)\ell(x).$$

X	Pr	Code I	Code II
0	$1/2$	000	0
1	$1/4$	001	10
2	$1/8$	010	110
3	$1/16$	011	1110
4	$1/64$	100	111100
5	$1/64$	101	111101
6	$1/64$	110	111110
7	$1/64$	111	111111

$$L_1(X) = 3$$

$$L_2(X) = 2$$

Source Coding Applications

- Magnetic recording: cassette, hard drive ...
- Speech compression
- Compact disk (CD)
- Image compression: JPEG

Source Coding: Set of codes

For $\mathcal{X} = \{1, 2, 3, 4\}$ and $\mathcal{D} = \{0, 1\}$, consider

x	$p(x)$	C_I	C_{II}	C_{III}	C_{IV}
1	1/2	0	0	10	0
2	1/4	0	1	00	10
3	1/8	1	00	11	110
4	1/8	10	11	110	111
$H(X)$	1.75	—	—	—	—
$E\ell(X)$	—	1.125	1.25	2.125	1.75

- Code efficiency = $H(X)/E[\ell(X)]$
- Which code is **best**? Would we prefer C_I or C_{II} ?

Consider C_I and decode string: 00001. It would come from 1, 2, 1, 2, 3 or 2, 1, 2, 1, 3 or 1, 1, 1, 1, 3, or etc.

Source Coding: Set of codes

For $\mathcal{X} = \{1, 2, 3, 4\}$ and $\mathcal{D} = \{0, 1\}$, consider

x	$p(x)$	C_I	C_{II}	C_{III}	C_{IV}
1	1/2	0	0	10	0
2	1/4	0	1	00	10
3	1/8	1	00	11	110
4	1/8	10	11	110	111
$H(X)$	1.75	—	—	—	—
$E\ell(X)$	—	1.125	1.25	2.125	1.75

- Code efficiency = $H(X)/E[\ell(X)]$
- Which code is **best**? Would we prefer C_I or C_{II} ?

Consider C_{II} and decode string: 0011. It could be either 1, 1, 2, 2 or 3, 4.

Source Coding: Set of codes

For $\mathcal{X} = \{1, 2, 3, 4\}$ and $\mathcal{D} = \{0, 1\}$, consider

x	$p(x)$	C_I	C_{II}	C_{III}	C_{IV}
1	1/2	0	0	10	0
2	1/4	0	1	00	10
3	1/8	1	00	11	110
4	1/8	10	11	110	111
$H(X)$	1.75	—	—	—	—
$E\ell(X)$	—	1.125	1.25	2.125	1.75

- Consider C_{III} . Can we decode 1100000000?

Yes. But if we only see a prefix, such as 11, we don't know **until we see more bits to the end.**

$$1100000000 = 3, 2, 2, 2, 2$$

$$11000000000 = 4, 2, 2, 2, 2$$

Source Coding: Set of codes

For $\mathcal{X} = \{1, 2, 3, 4\}$ and $\mathcal{D} = \{0, 1\}$, consider

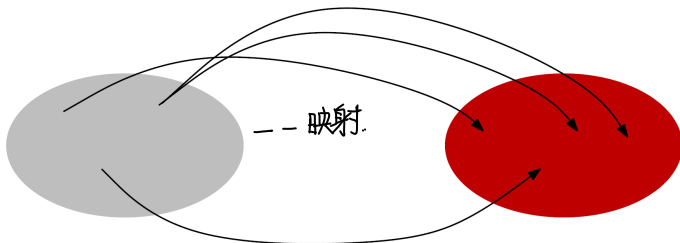
x	$p(x)$	C_I	C_{II}	C_{III}	C_{IV}
1	1/2	0	0	10	0
2	1/4	0	1	00	10
3	1/8	1	00	11	110
4	1/8	10	11	110	111
$H(X)$	1.75	—	—	—	—
$E\ell(X)$	—	1.125	1.25	2.125	1.75

- Consider C_{IV} . This code seems at least feasible (since $E[\ell] \geq H$). Decoding seems **easy**: (e.g., $111110100 = 111, 110, 10, 0 = 4, 3, 2, 1$).

Definition (Nonsingular Code)

A code C is called *nonsingular* if every realization of \mathcal{X} maps onto a difference codeword in \mathcal{D}^* , i.e.,

$$x \neq x' \Rightarrow C(x) \neq C(x').$$



Source Coding: Code types

Definition (Nonsingular Code)

A code C is called *nonsingular* if every element of \mathcal{X} maps onto a difference string in \mathcal{D}^* , i.e.,

$$x \neq x' \Rightarrow C(x) \neq C(x').$$

→ 不能区分编码方式的码。

x	$p(x)$	C_I	C_{II}	C_{III}	C_{IV}
1	1/2	0	0	10	0
2	1/4	0	1	00	10
3	1/8	1	00	11	110
4	1/8	10	11	110	111
$H(X)$	1.75	—	—	—	—
$E\ell(X)$	—	1.125	1.25	2.125	1.75

C_I is singular.

Source Coding: Code types

Definition (Code Extension)

The *extension* of a code $C: \mathcal{X} \rightarrow \mathcal{D}^*$ is defined by

$$C(x_1 x_2 \cdots x_n) = C(x_1) C(x_2) \cdots C(x_n).$$

Definition (Unique Decodable Code)

区分编码方式的好坏。
不能区分解码的好坏。

A code is called *uniquely decodable* if its extension is nonsingular.

$$x_1 x_2 \cdots x_m \neq x'_1 x'_2 \cdots x'_n \Rightarrow C(x_1 x_2 \cdots x_m) \neq C(x'_1 x'_2 \cdots x'_n)$$

Source Coding: Code types

Definition (Unique Decodable Code)

A code is called *uniquely decodable* if its extension is nonsingular.

C_{II}^* is **singular**. ($C(1, 1) = C(3) = 00$)

x	$p(x)$	C_I	C_{II}	C_{III}	C_{IV}
1	1/2	0	0	10	0
2	1/4	0	1	00	10
3	1/8	1	00	11	110
4	1/8	10	11	110	111
$H(X)$	1.75	—	—	—	—
$El(X)$	—	1.125	1.25	2.125	1.75

C_I is singular.

C_{II} is **NOT** u.d..

Source Coding: Code types

Definition (Unique Decodable Code)

A code is called *uniquely decodable* if its extension is nonsingular.

C_{III} is **uniquely decodable**.

x	$p(x)$	C_I	C_{II}	C_{III}	C_{IV}
1	1/2	0	0	10	0
2	1/4	0	1	00	10
3	1/8	1	00	11	110
4	1/8	10	11	110	111
$H(X)$	1.75	—	—	—	—
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C_I is singular.

C_{II} is **NOT** u.d..

Source Coding: Code types

Definition (Unique Decodable Code)

A code is called *uniquely decodable* if its extension is nonsingular.

$$1100000000 = 3, 2, 2, 2, 2$$

$$11000000000 = 4, 2, 2, 2, 2$$

To know the source, we
have to **wait until the end!**

x	$p(x)$	C_I	C_{II}	C_{III}	C_{IV}
1	1/2	0	0	10	0
2	1/4	0	1	00	10
3	1/8	1	00	11	110
4	1/8	10	11	110	111
$H(X)$	1.75	—	—	—	—
$El(X)$	—	1.125	1.25	2.125	1.75

C_I is singular.

C_{II} is **NOT** u.d..

Source Coding: Code types

Definition (Prefix Code) 一区分解码式的好坏

A code C is called a **prefix code** (a.k.a. **instantaneous**) iff no codeword of C is a prefix of any other codeword of C .

x	$p(x)$	C_I	C_{II}	C_{III}	C_{IV}
1	1/2	0	0	10	0
2	1/4	0	1	00	10
3	1/8	1	00	11	110
4	1/8	10	11	110	111
$H(X)$	1.75	—	—	—	—
$El(X)$	—	1.125	1.25	2.125	1.75

C_I is singular.

C_{II} is NOT u.d..

C_{III} is NOT prefix.

C_{IV} is prefix.

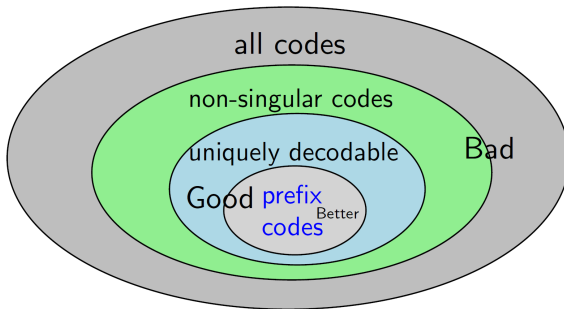
Source Coding: Code types

For $\mathcal{X} = \{1, 2, 3, 4\}$ and binary code, consider

x	$p(x)$	C_I	C_{II}	C_{III}	C_{IV}
1	1/2	0	0	10	0
2	1/4	0	1	00	10
3	1/8	1	00	11	110
4	1/8	10	11	110	111
$H(X)$	1.75	—	—	—	—
$E\ell(X)$	—	1.125	1.25	2.125	1.75

- C_I is singular.
- C_{II} is non-singular, but not uniquely decodable.
- C_{III} is non-singular, uniquely decodable, but NOT prefix.
- C_{IV} is non-singular, uniquely decodable, and prefix.

Source Coding: Classes of codes



- Goal: to find a **prefix code** with **minimum** expected length.

Kraft Inequality

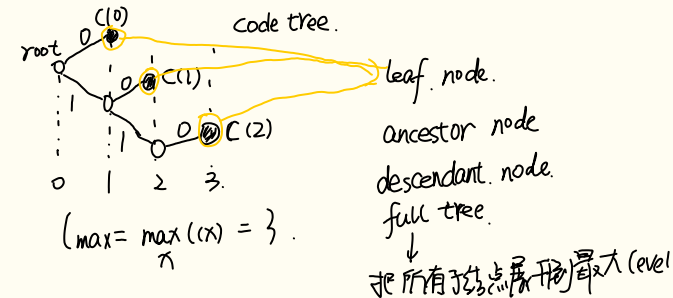
Theorem 5.2.1 (Kraft Inequality)

For any prefix code over an alphabet of size D , the codeword lengths l_1, l_2, \dots, l_m must satisfy the inequality

$$\sum_i D^{-l_i} \leq 1.$$

Conversely, given a set of codeword lengths that satisfy this inequality, there exists a prefix code with these codeword lengths.

x	$c(x)$	$x \in \{0, 1, 2\}$
0	0	
1	10	$D = \{0, 1\}$
2	110	



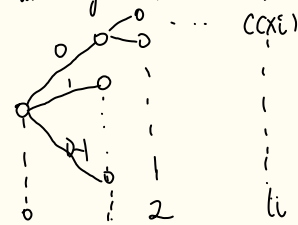
In prefix code.

codewords are leaf nodes of code tree

Proof:

Suppose there is a D -ary prefix code $\{c(x_1), c(x_2), \dots, c(x_n)\}$.

with lengths l_1, l_2, \dots, l_n , and code alphabet $D = \{0, 1, \dots, D-1\}$.



Expand the code tree to a full tree with l_{\max} levels.

② In the full tree, codeword $c(x_i)$ has $l_{\max} - l_i$ levels of descendants.

$\Rightarrow D^{l_{\max} - l_i}$ leaf nodes in the full tree are the descendants of $c(x_i)$.

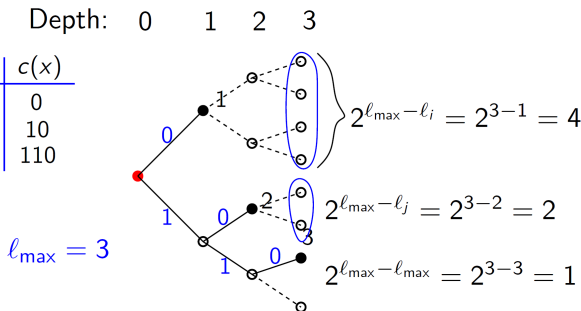
Kraft Inequality

Proof Idea. (A small example) To prove: A prefix code with lengths l_1, l_2, \dots, l_m , the inequality

$$\sum_i D^{-l_i} \leq 1 \text{ holds.}$$

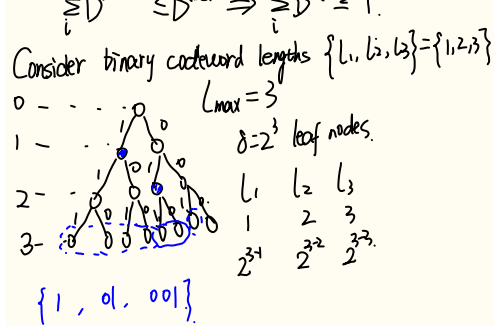
$D = \{0, 1\}$
 $\mathcal{X} = \{1, 2, 3\}$

x	c(x)
1	0
2	10
3	110



$$\sum_i 2^{-l_i} \leq 1 \Leftrightarrow \sum_i 2^{l_{\max} - l_i} \leq 2^{l_{\max}}$$

③ In full tree, leaf nodes of different codewords are not overlapped. Algorithm: l_1, l_2, \dots, l_m .
 前提: m 有限 \leftarrow suppose: $l_1 \leq l_2 \leq l_3 \leq \dots \leq l_m$



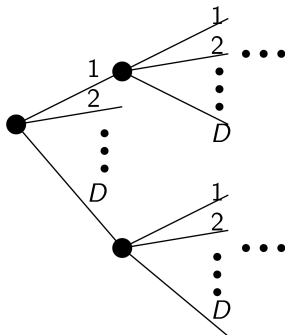
② let j be the minimum index of leaf nodes which has not been allocated, initially $j=0$.
 ③ for $i=1$ to m .
 pick up leaf nodes in $\{j, j+1, \dots, j+D^{l_{\max}-l_i}-1\}$
 find the node N_i , which is the closest common ancestor of above leaf nodes. $j = j + D^{l_{\max}-l_i}$
 end.
 output = $\{x_1, x_2, \dots, x_m\} \Rightarrow$ prefix code.

例: $m=\infty$,
 Poisson Distribution.
 $\Pr[X=k] = \frac{\lambda^k}{k!} e^{-\lambda}, k=0, 1, 2, \dots$
 $\mathcal{X} = \{0, 1, 2, 3, \dots\} \quad |\mathcal{X}| = \infty$

Kraft Inequality

Proof. (in general)

- Represent the set of prefix codes on a D -ary tree:

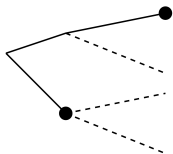


- Codewords correspond to leaves
- Path from root to each leaf determines a codeword
- **Prefix condition:** won't get to a codeword until we get to a leaf (no descendants of codewords are codewords)

Kraft Inequality

Proof. (in general)

- $\ell_{\max} = \max_i(\ell_i)$ is the length of the longest codeword.
- We can expand the full-tree down to depth ℓ_{\max} :



The nodes at the level ℓ_{\max} are either

- 1 codewords
 - 2 descendants of codewords
 - 3 neither
- Consider a codeword i at depth ℓ_i in tree
 - There are $D^{\ell_{\max} - \ell_i}$ descendants in the tree at depth ℓ_{\max}
 - Descendants of code i are **disjoint** from decedents of code j (prefix free condition)

Kraft Inequality

Proof. (in general)

- All the above implies:

$$\sum_i D^{\ell_{\max} - \ell_i} \leq D^{\ell_{\max}} \Rightarrow \sum_i D^{-\ell_i} \leq 1$$

- **Conversely:** given codewords lengths $\ell_1, \ell_2, \dots, \ell_m$ satisfying Kraft inequality, try to construct a **prefix code**.

Kraft Inequality

Proof. (in general)

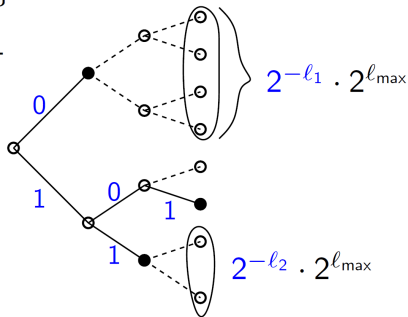
- **Conversely:** given codewords lengths $\ell_1, \ell_2, \dots, \ell_m$ satisfying Kraft inequality, try to construct a prefix code.

$$\{\ell_1, \ell_2, \ell_3\} = \{1, 2, 3\}$$

$$2^{-1} + 2^{-2} + 2^{-3} < 1$$

x	$c(x)$
1	0
2	11
3	101

C is prefix.



Reading & Homework

Reading : 5.1, 5.2

Homework : Problems 5.1, 5.3, 5.18, 5.37