

Introduction to Information Theory

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Plan

- Motivations and introduction
- Source model (DMS)
- Information content
- Entropy
- Information rate
- Based on the book:
“Information Theory, Inference, and Learning Algorithms”. David J.C. MacKay

Motivations

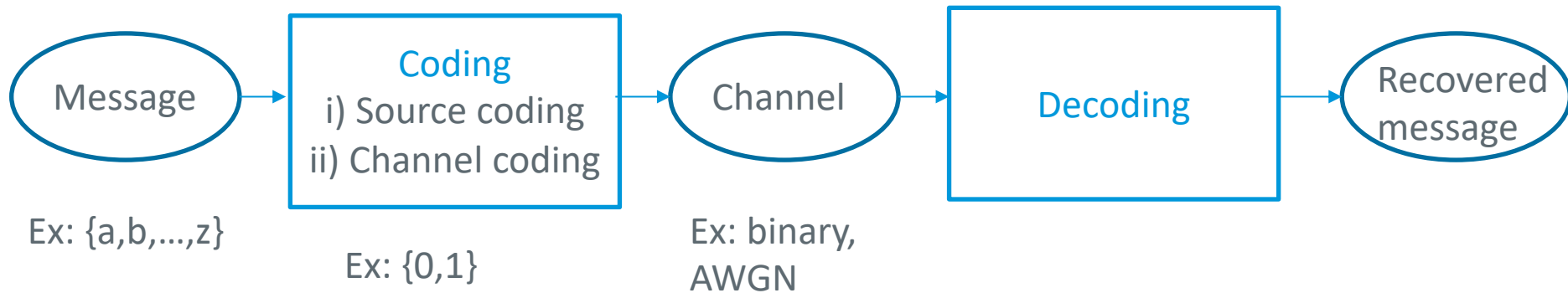
- Noisy communication channels
 - Noise: limits impossible to be overcome?
- Claude Shannon
 - Possibility to transmit information with prob. of error as small as desired provided a **suitable code** is used
- Important concepts and tools:
 - Quantification of the information?
 - Random noise/errors: probability/statistics



Claude Shannon

- (April 30, 1916 – February 24, 2001) was an American mathematician and electronic engineer known as “the father of information theory”
- Shannon is best known for his 1948 paper “*A Mathematical Theory of Communication*” in which he created the field of Information Theory, but he had many other important contributions in diverse areas.

Typical communication system



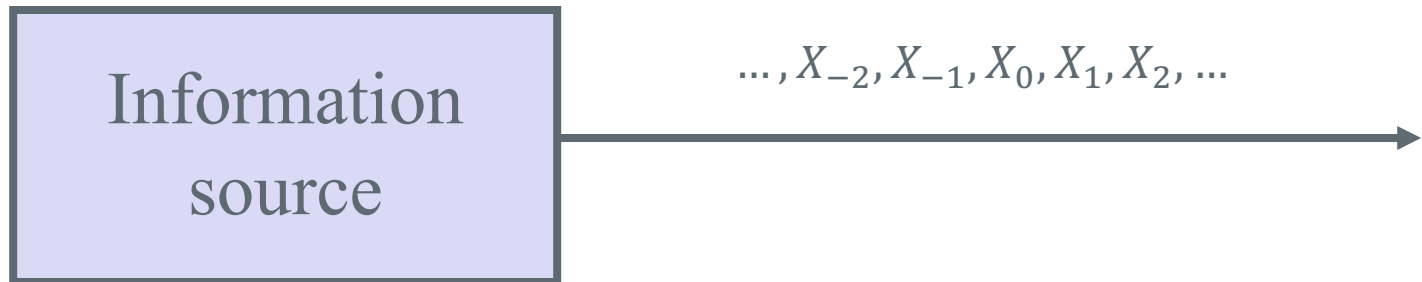
- Coding:
 - Represent the message as efficiently as possible
 - Make the make robust to channel noise
- Joint source/channel coding (~1990) (not covered here)
- 2 types of sources
 - Discrete/continuous

Source/channel coding

- Different objectives
- Source coding
 - Reduce redundancy
 - Shorten the message length
 - Efficient use of bandwidth
- Channel coding
 - Add redundancy
 - Reduce decoding errors
 - Depends on the transmission environment
- Shannon showed that this can be handled separately

Example of source model

- Discrete memoryless source (DMS)
 - A mathematical model for information source: a discrete-time random process $\{X_i\}_{i=-\infty}^{\infty}$



- The simplest model for the information source is the *discrete memoryless source* (DMS). A DMS is a discrete time, discrete-amplitude random process in which **all X_i are generated independently and with the same distribution.**

Discrete memoryless source

- X takes values from the alphabet $\mathcal{A} = \{a_1, a_2, \dots, a_N\}$ with probabilities $p_i = \mathbb{P}(X = a_i)$
- A full description of the DMS is given by the set \mathcal{A} (alphabet), and the probabilities $\{p_i\}_{i=1}^N$

Examples

- $\mathcal{A} = \{0,1\}$ (binary source)
- $\mathbb{P}(X = 0) = p, \mathbb{P}(X = 1) = 1 - p$
- Transmitted data: 001011000100 ...
- $\mathcal{A} = \{00,01,10\}$
- $\mathbb{P}(X = 00) = 0.3, \mathbb{P}(X = 01) = 0.5, \mathbb{P}(X = 10) = 0.2$
- Transmitted data: 0010~~11~~000100 ...

Impossible: not in alphabet

Memoryless source

- Property: all the symbols are mutually independent

$$\mathbb{P}(X_1, X_2, \dots, X_N) = \mathbb{P}(X_1)\mathbb{P}(X_2) \dots \mathbb{P}(X_N)$$

- Other source model: Markovian source
 - Each symbol depends of the previous one
$$\mathbb{P}(X_1, X_2, \dots, X_N) = \mathbb{P}(X_1)\mathbb{P}(X_2|X_1) \dots \mathbb{P}(X_N|X_{N-1})$$
 - Not covered in this course

Different source models

- Zero-order approximation (symbols independent and equiprobable)
XFOML RXKHRJFFJUJ ZLPWCFWKCYJ
- First-order approximation (symbols independent but with frequencies of English text)
OCRO HLI RGWR NMIELWIS EU LL
- Second-order approximation (digram structure as in English)
ON IE ANTSOUTINYS ARE T INCTORE ST BE
- Third-order approximation (trigram structure as in English)
IN NOT IST LAT WHEY CRATICT FROURE BIRS GROCID

Back to the binary source

- $\mathcal{A} = \{0,1\}$ (binary source)
- $\mathbb{P}(X = 0) = p, \mathbb{P}(X = 1) = 1 - p$
- Transmitted data: 001011000100 ...
- Binary symmetric channel



$$\mathbb{P}(Y = 0|X = 0) = 1 - f, \mathbb{P}(Y = 1|X = 1) = 1 - f$$

$$\mathbb{P}(Y = 1|X = 0) = f, \mathbb{P}(Y = 0|X = 1) = f$$

Back to the binary source

- f : Probability of error of a symbol (does not depend on the actual symbol)
- Problem: for $f = 0.1$, 10% of the symbols are corrupted!

$$\mathbb{P}(Y \neq X) = \mathbb{P}(Y = 1|X = 0)\mathbb{P}(X = 0) + \mathbb{P}(Y = 0|X = 1)\mathbb{P}(X = 1) = fp + f(1 - p) = f$$

- How can this be improved?

Example: repetition code

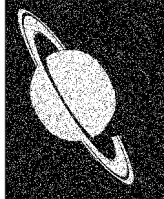
- “Naïve” approach: repeated code
 - If we send the data twice, we can detect some errors, but not correct them all
- $0 \rightarrow 00$, if 01 is received should we decide 0 or 1?
- R_N repetition: demo1.m

Repetition code

Original Image



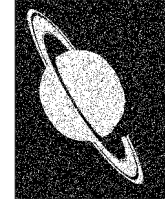
Received Image (1)



Received Image (2)



Received Image (3)



Decoded Image



Repetition code

- $\mathcal{A} = \{0,1\}$ (binary source)
- $\mathbb{P}(X = 0) = 0.5, \mathbb{P}(X = 1) = 0.5$ ($p = 0.5$)
- $f=0.1$
- Using the R_3 coding/decoding strategy, by how much can we improve the error?
- Demo1.m (+ tutorial)