COMP2610 / COMP6261 Information Theory Lecture 17: Noisy Channels

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Acknowledgement: These slides were originally developed by Professor Robert C. Williamson.



Announcements

- Assignment 3 is now available!
 - 1 Due Date: Monday 24 October 2021, 9:00 a.m.
 - 2 Weighting: 20% of the course mark.
- Past Exam papers are available in Wattle

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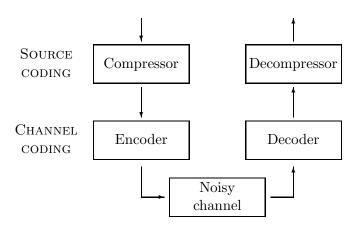
Communication over Noisy Channels: Big Picture

Noisy Channels: Formally

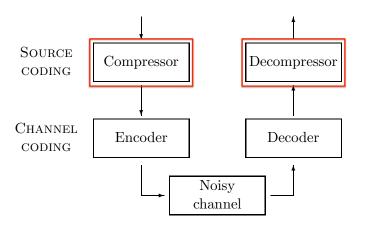
Examples of Channels

Probability of Error

The Big Picture



The Big Picture

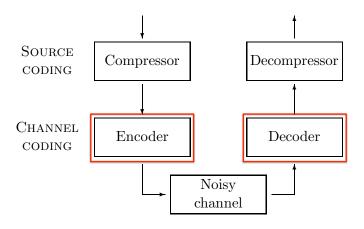


Concept: Expected code length

Theorem: Source coding theorem

Algorithms: { Huffman, Arithmetic } codes

The Big Picture



Concept: Channel capacity

Theorem: Channel coding theorem

Algorithms: Repetition codes, Hamming codes

Communication over Noisy Channels

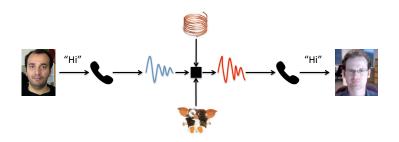
A channel is some medium for transmitting messages

A noisy channel is a channel that potentially introduces errors in the sender's message

The Problem of Communication

"The fundamental problem of communication is that of reproducing at one point either exactly or approximately a message selected at another point." (Claude Shannon, 1948)

Example: Telephone Network



Source : Aditya

Encoder: Telephone handset

Channel: Analogue telephone line

Decoder: Telephone handset

Destination: Mark

Key Questions

How do we model noisy communication abstractly?

What are the theoretical limits of noise correction?

What are the practical approaches to noise correction?

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3 Examples of Channels

Probability of Error

Suppose we have some set $S = \{1, 2, ..., S\}$ of possible messages

- e.g. codewords from Huffman coding on some ensemble
- Sender and receiver agree on what these are

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When communicating over a channel, the sender must encode messages into some input alphabet $\mathcal X$

The receiver then receives some (possibly corrupted) element from an output alphabet $\ensuremath{\mathcal{Y}}$

- Simple case: $\mathcal{X} = \mathcal{Y} = \{0, 1\}$
- The bit the sender transmits may not be what the receiver sees

Formally, the sender encodes messages via

 $\mathtt{enc} \colon \mathcal{S} \to \mathcal{X}^{\textit{N}}$

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enc:
$$\mathcal{S} o \mathcal{X}^{ extsf{N}}$$

The receiver then decodes messages via

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Isn't the compressor already "encoding" a message?

Yes, but we might want to add something for noise tolerance

Might have $\mathcal{X} \neq \mathcal{Y}$

- e.g. if we allow a special "erased" symbol
- N > 1 can be thought of as multiple uses of a channel
 - e.g. use a bitstring of length 4 to represent messages

Channels: Informally

A discrete channel Q will:

- ullet accept an **input** from ${\mathcal X}$
- ullet produce an **output** from ${\mathcal Y}$

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- ullet produce an **output** from ${\mathcal Y}$

There is a probability of observing various outputs, given an input

- This represents some inherent noise
- Noise could depend on the input

Channels: Formally

A discrete channel Q consists of:

- an input alphabet $\mathcal{X} = \{a_1, \dots, a_l\}$
- an output alphabet $\mathcal{Y} = \{b_1, \dots, b_J\}$
- transition probabilities P(y|x).

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The channel Q can be expressed as a matrix

$$Q_{j,i}=P(y=b_j|x=a_i)$$

This represents the probability of observing b_j given that we transmit a_i

Channels: Example

Example: A channel Q with inputs $\mathcal{X} = \{a_1, a_2, a_3\}$, outputs $\mathcal{Y} = \{b_1, b_2\}$, and transition probabilities expressed by the matrix

$$Q = \begin{bmatrix} 0.8 & 0.5 & 0.2 \\ 0.2 & 0.5 & 0.8 \end{bmatrix}$$

So
$$P(b_1|a_1) = 0.8 = P(b_2|a_3)$$
 and $P(b_1|a_2) = P(b_2|a_2) = 0.5$.

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We arrange the inputs along the columns, and outputs along the rows

Actual details of alphabet are abstracted away

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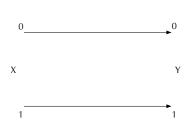
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Probability of Error

The Binary Noiseless Channel

One of the simplest channels is the **Binary Noiseless Channel**. The received symbol is always equal to the transmitted symbol – there is no probability of error, hence *noiseless*.

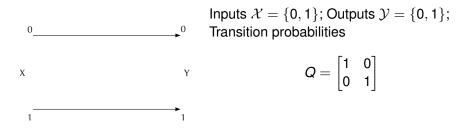


Inputs $\mathcal{X} = \{0,1\}$; Outputs $\mathcal{Y} = \{0,1\}$; Transition probabilities

$$Q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

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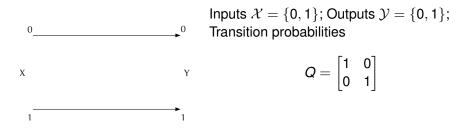


What was transmitted over the channel if 0000 1111 was received?

$$\xrightarrow{Q}$$
 0000 1111

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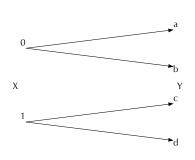


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The Noisy Non-overlapping Channel

Even if there is some uncertainty about the output given the input, it may still be possible to perfectly infer what was transmitted.

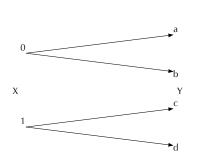


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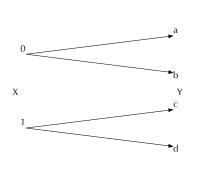
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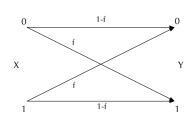
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Each symbol sent across a **binary symmetric channel** has a chance of being "flipped" to its counterpart $(0 \rightarrow 1; 1 \rightarrow 0)$

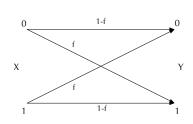


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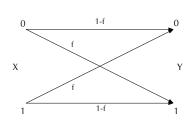
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What was most likely transmitted over the channel if 0010 1001 was received, assuming f = 0.1 and P(x = 0) = P(x = 1) = 0.5?

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Inferring the Input

Suppose we know the P(x) — the probability x is transmitted.

Given a particular output $y \in \mathcal{Y}$ received over a channel Q, how likely was it that input $x \in \mathcal{X}$ was transmitted?

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$$P(x|y) = \frac{P(y|x)P(x)}{P(y)} = \frac{P(y|x)P(x)}{\sum_{x' \in \mathcal{X}} P(y|x')P(x')}$$

Inferring the Input: Example

Suppose P(x = 0) = P(x = 1) = 0.5. What are the probability that a x = 0 was transmitted over a *binary symmetric channel Q* with f = 0.1 given that a y = 0 was received?

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Similarly, P(x = 1|y = 1) = 0.9.

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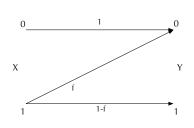
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What if
$$P(x = 0) = 0.01$$
?

$$P(x=0|y=0) = \frac{0.9 \times 0.01}{0.9 \times 0.01 + 0.1 \times 0.99} \approx 0.0833.$$

The Z Channel

Symbols may be corrupted over the channel asymmetrically.

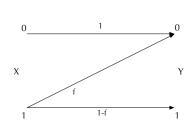


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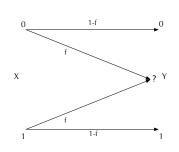
Inferring the input: Clearly P(x = 1|y = 1) = 1 but

$$P(x=0|y=0) = \frac{P(y=0|x=0)P(x=0)}{\sum_{x'\in\mathcal{X}} P(y=0|x')P(x')} = \frac{P(x=0)}{P(x=0) + fP(x=1)}$$

So $P(x=0|y=0) \rightarrow 1$ as $f \rightarrow 0$, and goes to P(x=0) as $f \rightarrow 1$

The Binary Erasure Channel

We can model a channel which "erases" bits by letting one of the output symbols be the symbol '?' with associated probability *f*. The receiver knows which bits are erased.



Inputs $\mathcal{X} = \{0, 1\}$; Outputs $\mathcal{Y} = \{0, ?, 1\}$; Transition probabilities

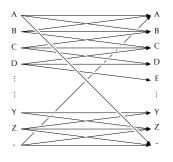
$$Q = \begin{bmatrix} 1 - f & 0 \\ f & f \\ 0 & 1 - f \end{bmatrix}$$

Example:

$$0000\ 1111 \xrightarrow{Q} 00?0\ ?11?$$

This channel simulates a noisy "typewriter". Inputs and outputs are 26 letters A through Z plus space. With probability $\frac{1}{3}$, each letter is either: unchanged; changed to the next letter, changed to the previous letter.

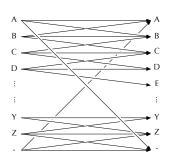
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Inputs $\mathcal{X} = \{\mathtt{A},\mathtt{B},\ldots,\mathtt{Z},_\};$ Outputs $\mathcal{Y} = \{\mathtt{A},\mathtt{B},\ldots,\mathtt{Z},_\};$ Transition probabilities

$$Q = \begin{bmatrix} \frac{1}{3} & \frac{1}{3} & 0 & 0 & \cdots & 0 & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{3} & \frac{1}{3} & 0 & \cdots & 0 & 0 \\ 0 & \frac{1}{3} & \frac{1}{3} & \frac{1}{3} & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \frac{1}{3} & 0 & 0 & & \cdots & \frac{1}{3} & \frac{1}{3} \end{bmatrix}$$

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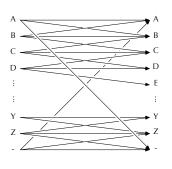


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The transition matrix for this channel has a diagonal structure: all of the probability mass is concentrated around the diagonal.

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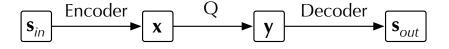
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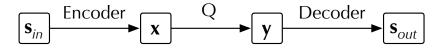
Communicating over Noisy Channels

Suppose we know we have to communicate over some channel Q and we want build an *encoder/decoder* pair to reliably send a message s over Q.



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Suppose we know we have to communicate over some channel Q and we want build an *encoder/decoder* pair to reliably send a message \mathbf{s} over Q.



Reliability is measured via **probability of error** — that is, the probability of incorrectly decoding \mathbf{s}_{out} given \mathbf{s}_{in} as input:

$$P(\mathbf{s}_{out} \neq \mathbf{s}_{in}) = \sum_{\mathbf{s}} P(\mathbf{s}_{out} \neq \mathbf{s}_{in} | \mathbf{s}_{in} = \mathbf{s}) P(\mathbf{s}_{in} = \mathbf{s})$$

Let
$$\mathcal{S} = \{a,b\}$$
 and $\mathcal{X} = \mathcal{Y} = \{0,1\}$

Assume encoder: $a \rightarrow 0$; $b \rightarrow 1$, decoder: $0 \rightarrow a$; $1 \rightarrow b$.

Consider binary symmetric Q,

$$Q = \begin{bmatrix} 1 - f & f \\ f & 1 - f \end{bmatrix}$$

with *f*= 0.1

$$P(\mathbf{s}_{in} \neq \mathbf{s}_{out}) = P(\mathbf{s}_{in} = \mathbf{a}, \mathbf{s}_{out} = \mathbf{b}) + P(\mathbf{s}_{in} = \mathbf{b}, \mathbf{s}_{out} = \mathbf{a})$$

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$$P(\mathbf{s}_{out} = a | \mathbf{s}_{in} = b) P(\mathbf{s}_{in} = b)$$

$$= P(y = 1 | x = 0) p_a + P(y = 0 | x = 1) p_b$$

$$= f$$

$$= 0.1$$

Suppose $\mathbf{s} \in \{a,b\}$ and we encode by $a \to 000$ and $b \to 111$. To decode we count the number of 1s and 0s and set all bits to the majority count to determine \mathbf{s}

$$\underbrace{000,001,010,100}_{A} \rightarrow \text{a} \quad \text{and} \quad \underbrace{111,110,101,011}_{B} \rightarrow \text{b}$$

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If the channel Q is binary symmetric with f = 0.1 again

$$P(\mathbf{s}_{in} \neq \mathbf{s}_{out}) = P(\mathbf{y} \in B|000) \, \rho_{a} + P(\mathbf{y} \in A|111) \, \rho_{b}$$

$$= [f^{3} + 3f^{2}(1 - f)] \rho_{a} + [f^{3} + 3f^{2}(1 - f)] \rho_{b}$$

$$= f^{3} + 3f^{2}(1 - f) = 0.028$$

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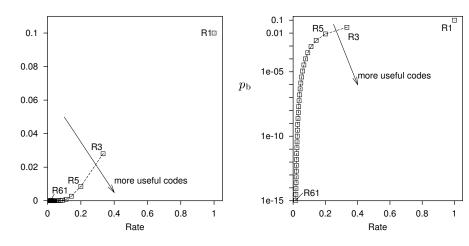
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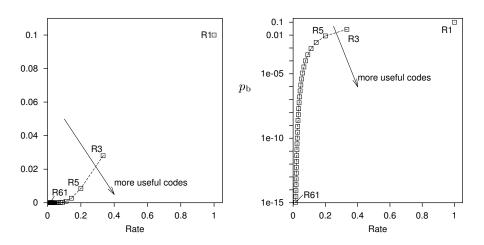
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$$= [f^{3} + 3f^{2}(1-f)] p_{a} + [f^{3} + 3f^{2}(1-f)] p_{b}$$

$$= f^{3} + 3f^{2}(1-f) = 0.028$$

So the *error* has dropped from 0.1 to 0.028 but so has the *rate*: from 1 symbol/bit to 1/3 symbol/bit.





Can we make the error arbitrarily small without the rate going to zero?

Summary and Reading

Main Points:

- Modelling Noisy Channels
 - Noiseless, Overlap, Symmetric, Z, Erasure
- Simple Coding via Repetition
 - Probability of Error vs Transmission Rate

Reading:

MacKay §9.1 - §9.5

Cover & Thomas §7.1 - §7.3