

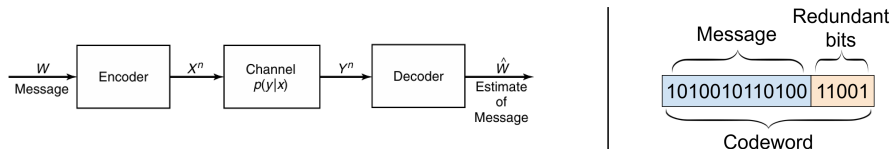
# LLM-Based Codes for Deletion Channels

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# Review of Deletion Channels



$\mathcal{X}$  = channel input alphabet     $\mathcal{Y}$  = channel output alphabet

Deletion Channel: Every symbol  $X_i$  in message  $X \in \mathcal{X}^n$  is dropped from the message with probability  $p$

Example:  $\mathcal{X} = \{0, 1\}$ ,  $\mathcal{Y} = \{0, 1\}$

$X = 1010$

$Y = 110$

# Shannon's Noisy-Channel Coding Theorem

Every channel has a capacity  $C = \sup_{p(X)} I(X; Y)$ .

## Theorem (Shannon's Noisy-Channel Coding Theorem)

*For any rate  $R < C$ , there exists a code that achieves  $R$  with arbitrarily small probability of error.*

Notes:

- ▶ Shannon did not tell us *how* to achieve this rate
- ▶ The capacity of the deletion channel is not known

# Project Vision

Use the English alphabet as input/output alphabet

$$\mathcal{X} = \{A, B, C, \dots, X, Y, Z, a, b, c, \dots, x, y, z\}$$
$$\mathcal{Y} = \{A, B, C, \dots, X, Y, Z, a, b, c, \dots, x, y, z\}$$

Exploit inherent redundancy:

“DIACS REUis th coolesresearch pogam ver ceatd!”

Use LLMs as a tool to recover text with deletions

# Is GPT Better Than You?

Deleted text ( $p = 0.30$ ):

"heelee wee on of three easvewed as bein inthe nnior Rdgers."

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Deleted text ( $p = 0.30$ ):

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Original:

"The Steelers were one of three teams viewed as being in the running for Rodgers."

# Is GPT Better Than You?

Deleted text ( $p = 0.30$ ):

“heelee wee on of three easviewed as bein inthe nnior Rdgers.”

Original:

“The Steelers were one of three teams viewed as being in the running for Rodgers.”

GPT-recovered:

“The Steelers were one of three teams viewed as being in the running for Rodgers.”

# Creating Datasets

Need for large amounts of plaintext

Method 1: Scraping recent news articles

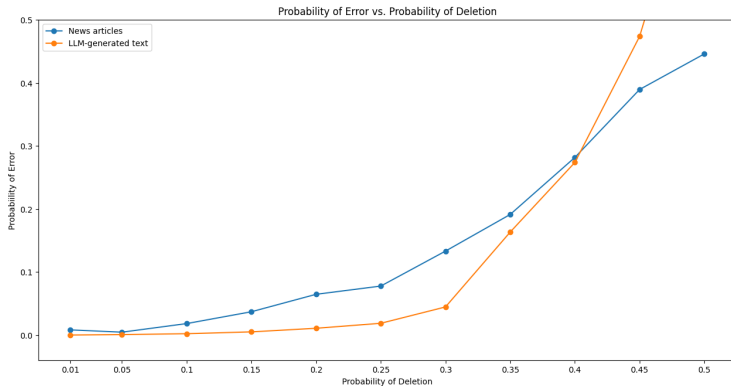
- ▶ “After Saturday’s 3-1 loss to Turkey...”
- ▶ “An international team led by Dr. Andy Rivkin...”
- ▶ “<b>What seafood should you avoid?</b>”

Method 2: Generating “clean” texts using LLMs

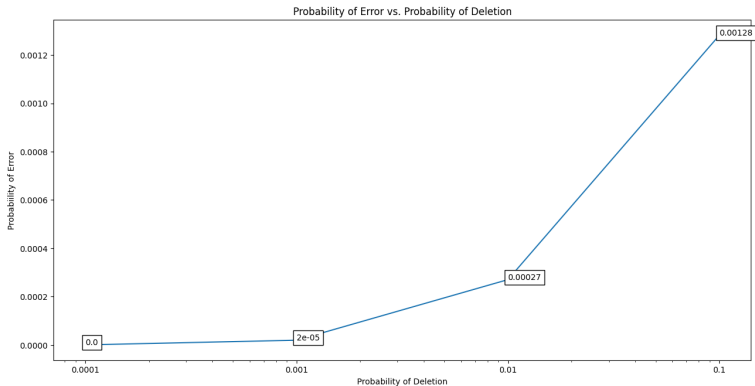
- ▶ “...divide cashflows into tranches...”
- ▶ “...crucial for companies operating loally...”



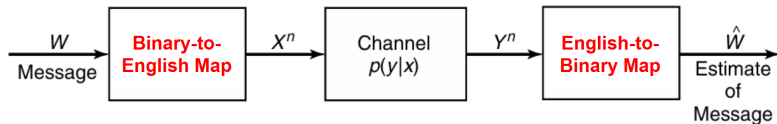
# GPT Performance on Datasets



# GPT Performance on Datasets



# Mapping Binary to a Language



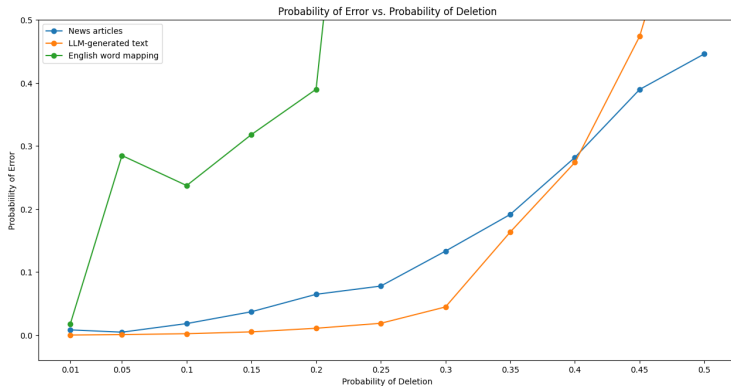
Idea 1: Map all  $2^b$  binary strings of length  $b$  to English *words*

- ▶ 0000 → apple
- ▶ 0001 → banana
- ▶ ...

Idea 2: Map all  $2^b$  binary strings of length  $b$  to English *sentences*

- ▶ 0000 → The sky is blue.
- ▶ 0001 → P equals NP.
- ▶ ...

# GPT Performance Using Mapping



# Next Steps



- ▶ Explore more efficient mappings from binary to some structured language
- ▶ Explore alternatives to edit distance (semantic distance)
- ▶ Automatic identification of parts of text that cannot be inferred from context
- ▶ Theoretical guarantees on performance

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# References

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