

On the Capacity of DNA-based Data Storage under Substitution Errors

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UC San Diego



May 9th, 2022

Outline

- Channel Model
- Related Work
- Preliminaries
- Channel Capacity
- Summary & Outlook

Data Storage in DNA



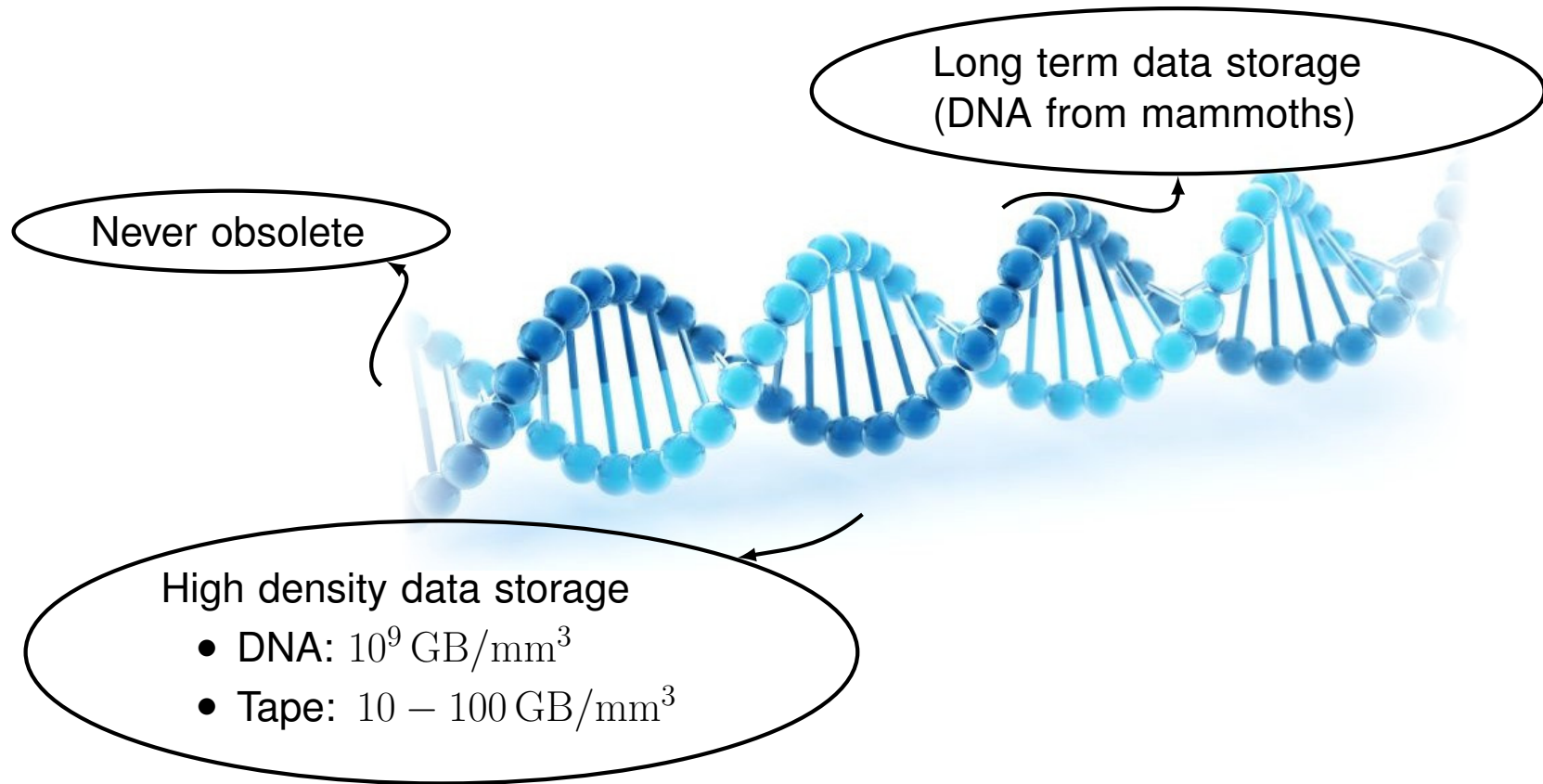
High density data storage

- DNA: 10^9 GB/mm³
- Tape: 10 – 100 GB/mm³

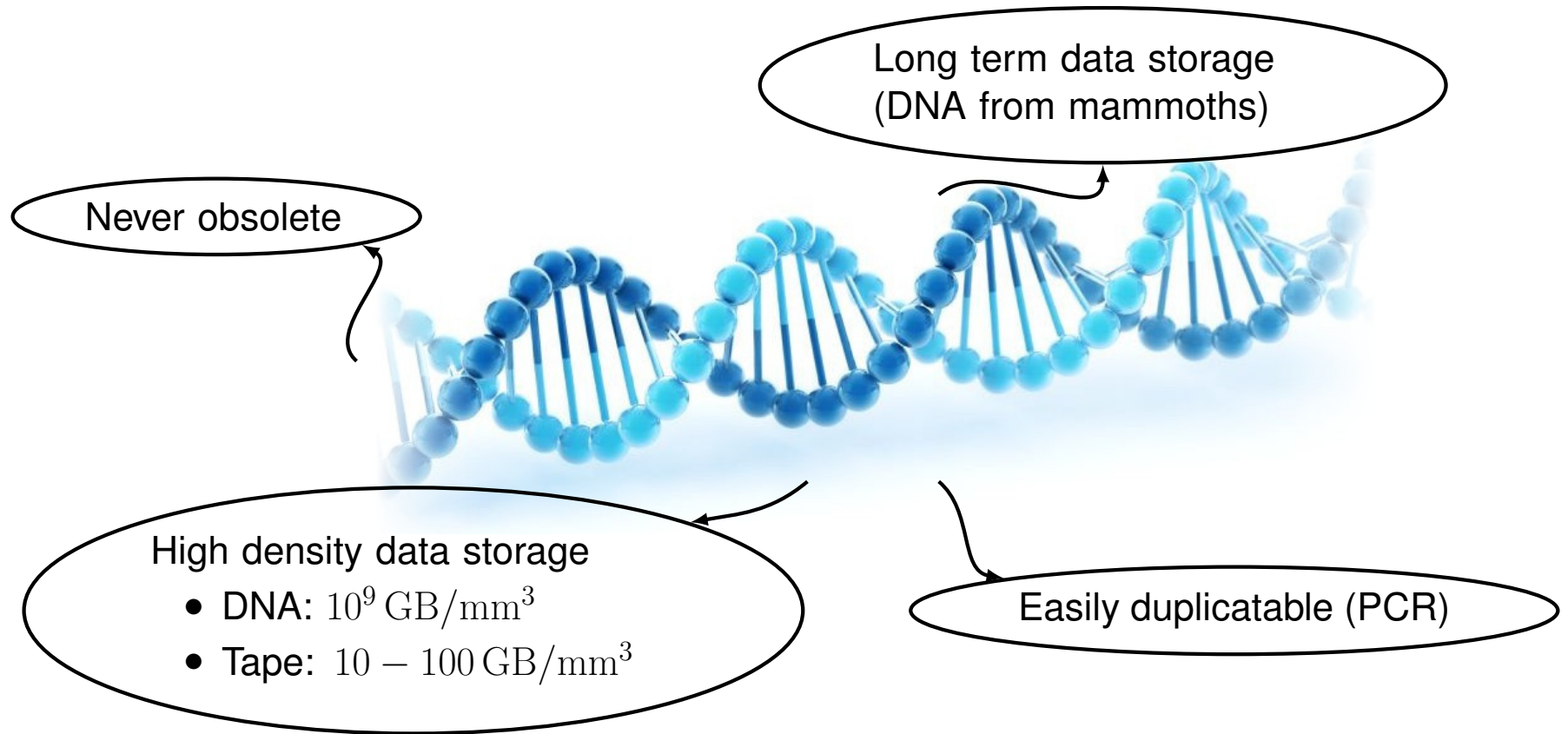
Data Storage in DNA



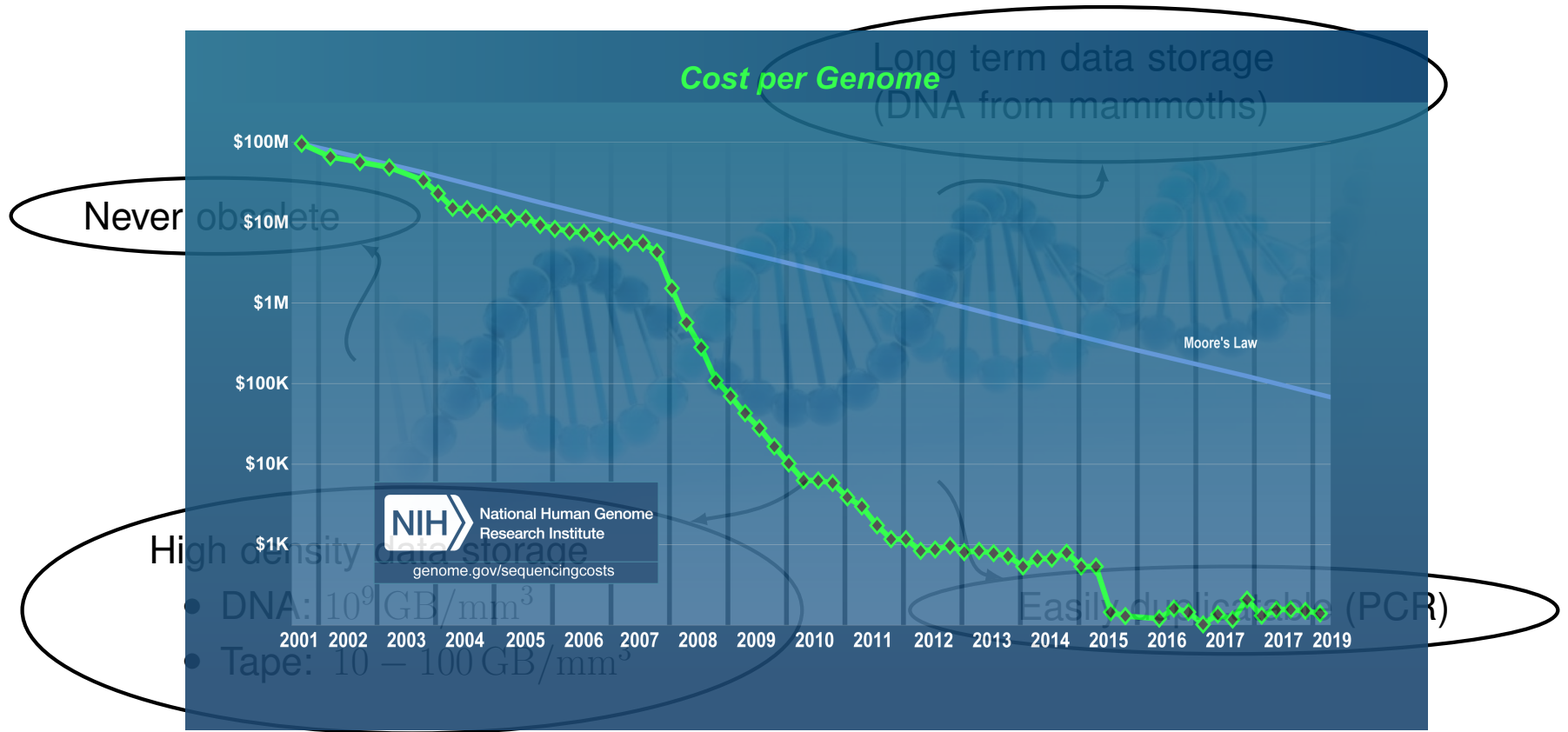
Data Storage in DNA



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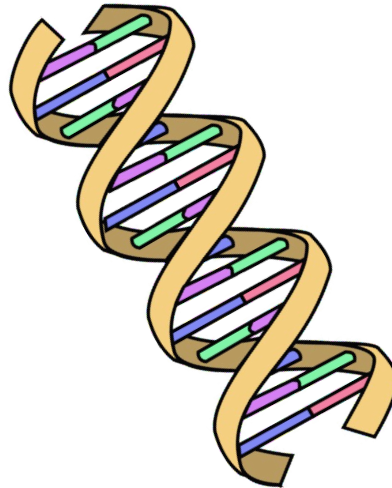
Data Storage in DNA

User Binary Data

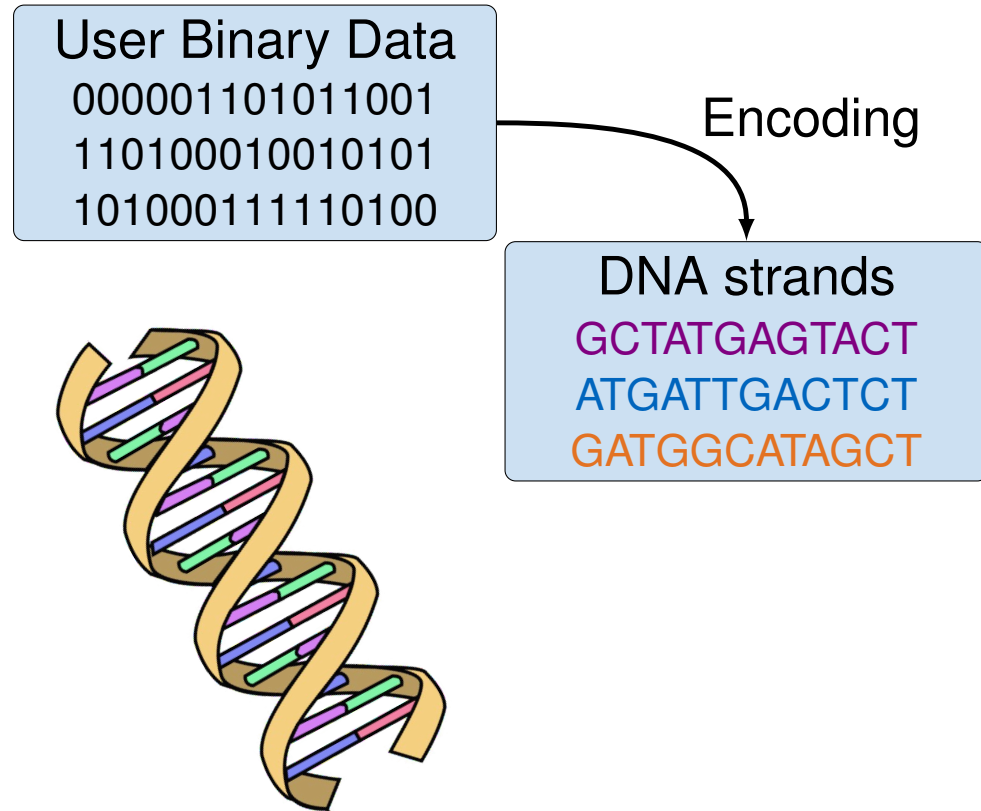
000001101011001

110100010010101

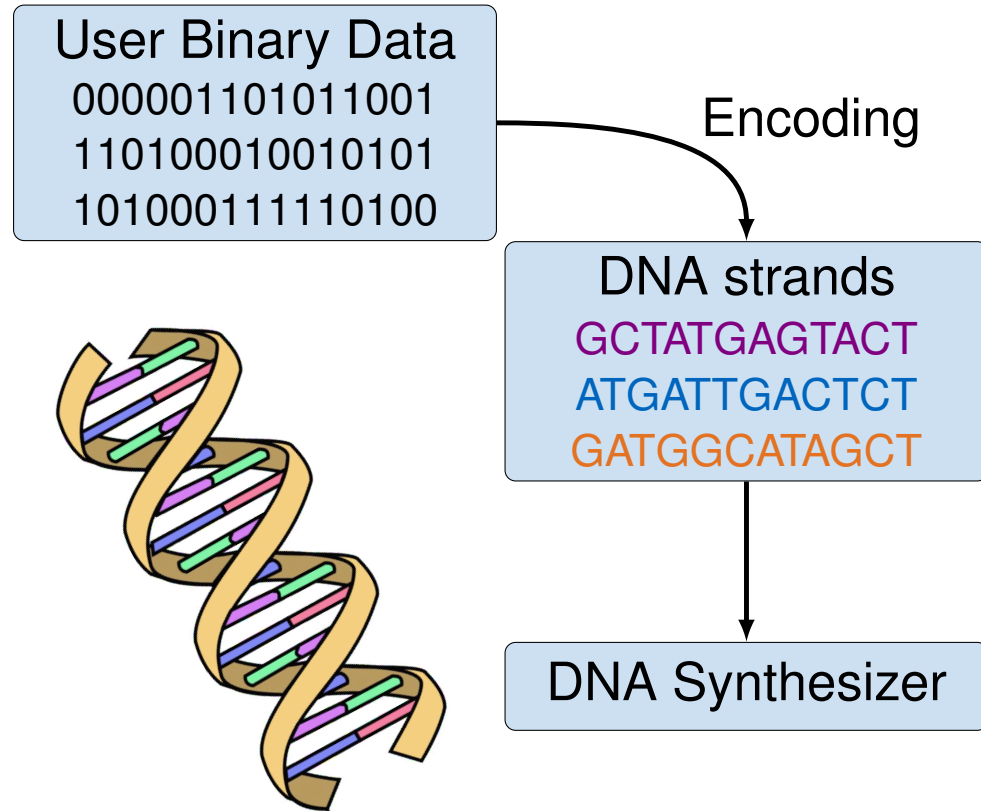
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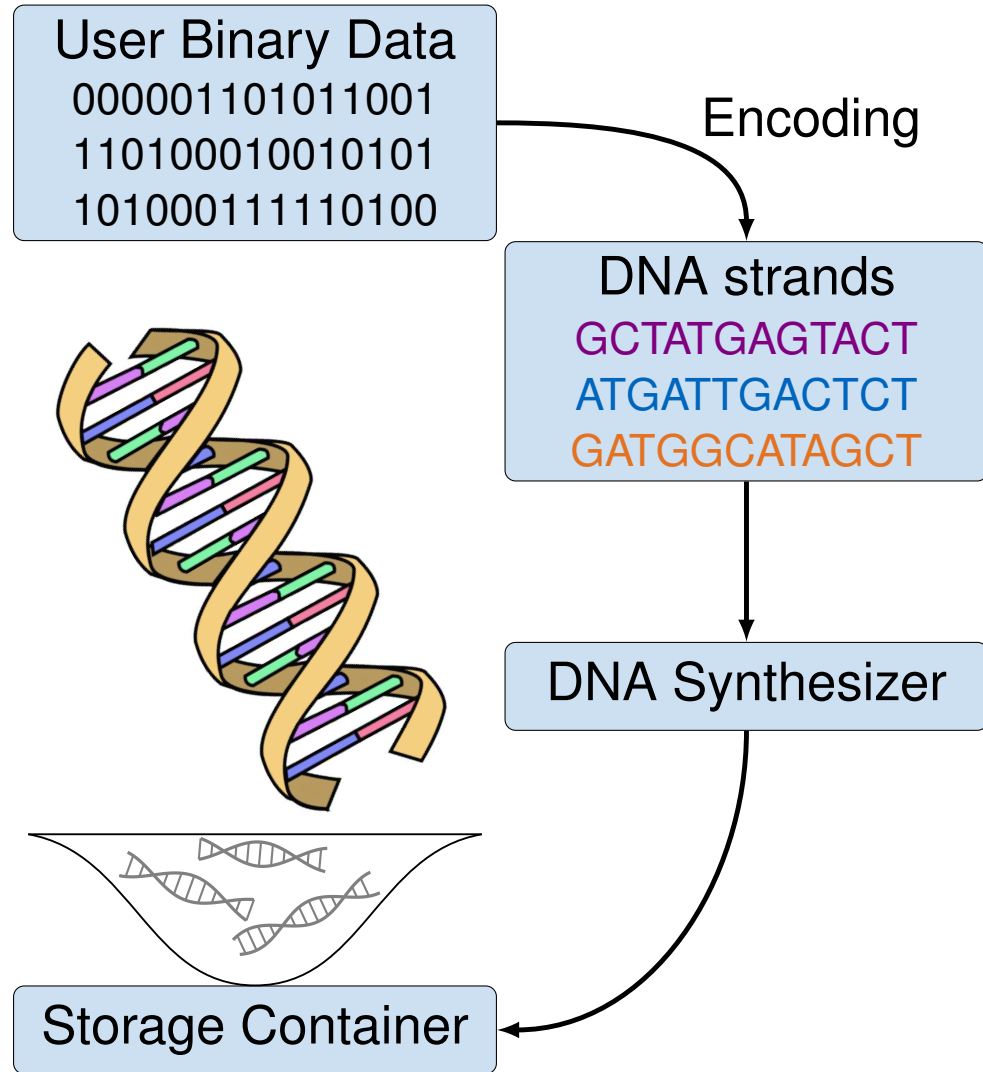
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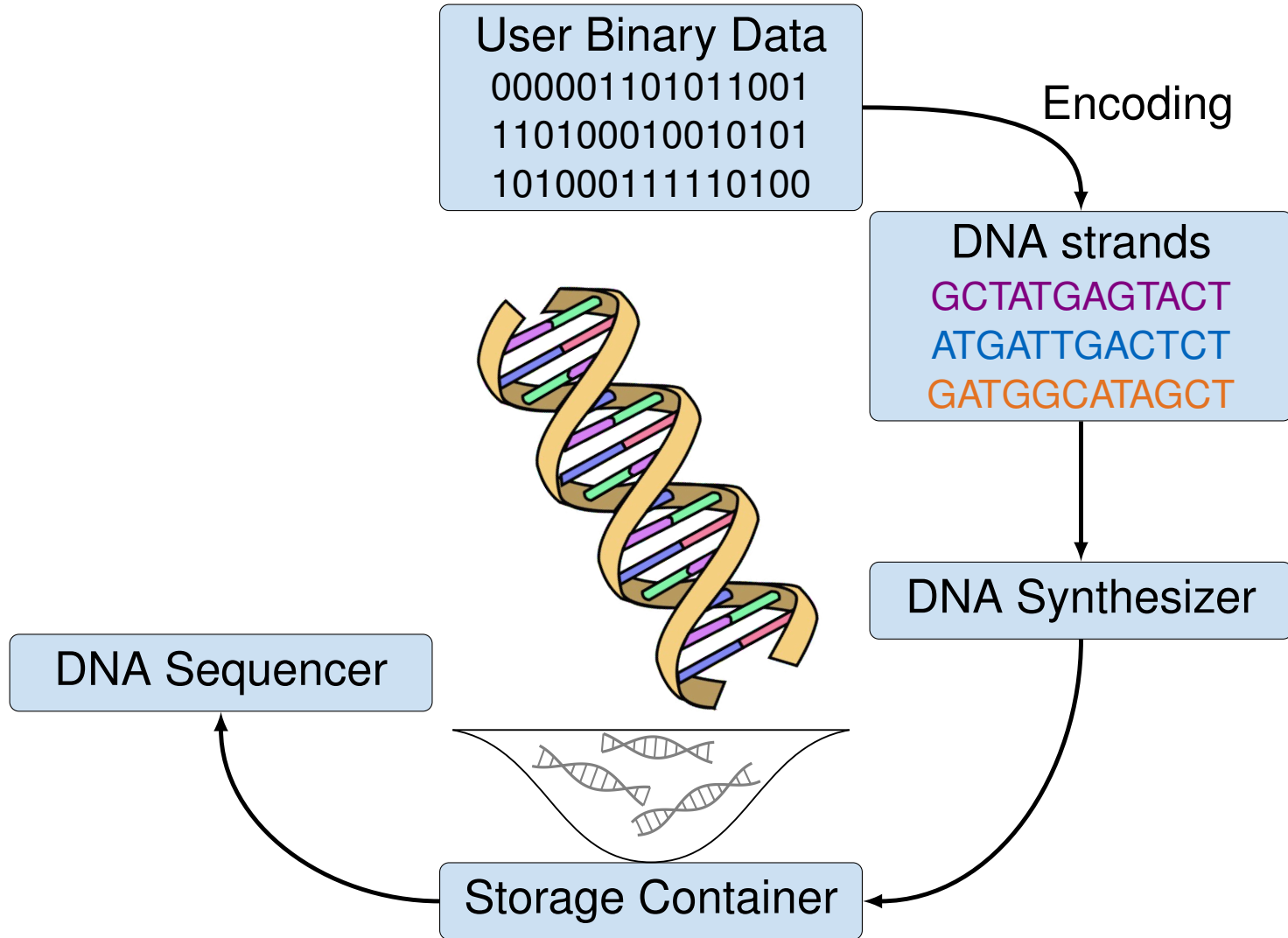
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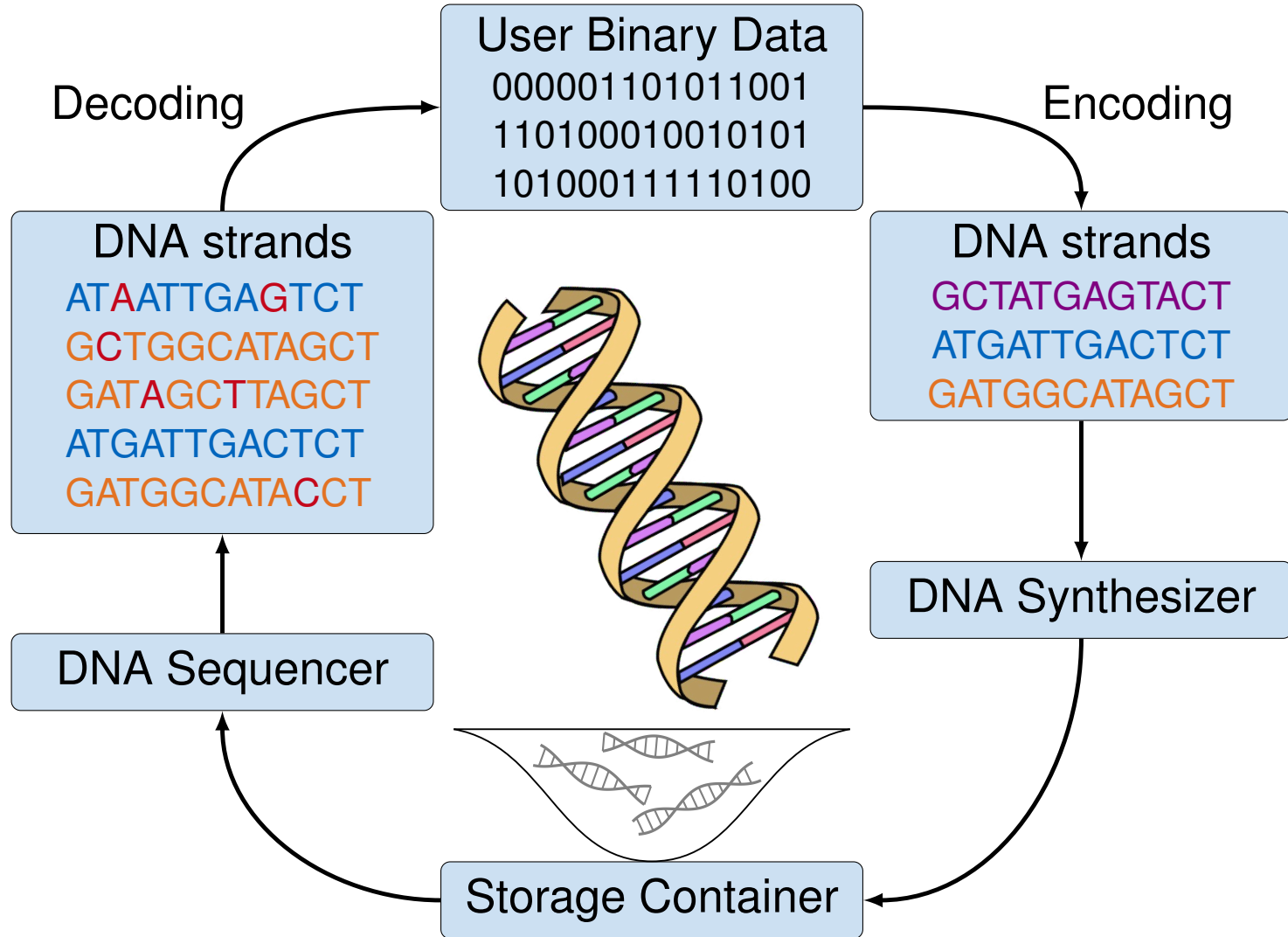
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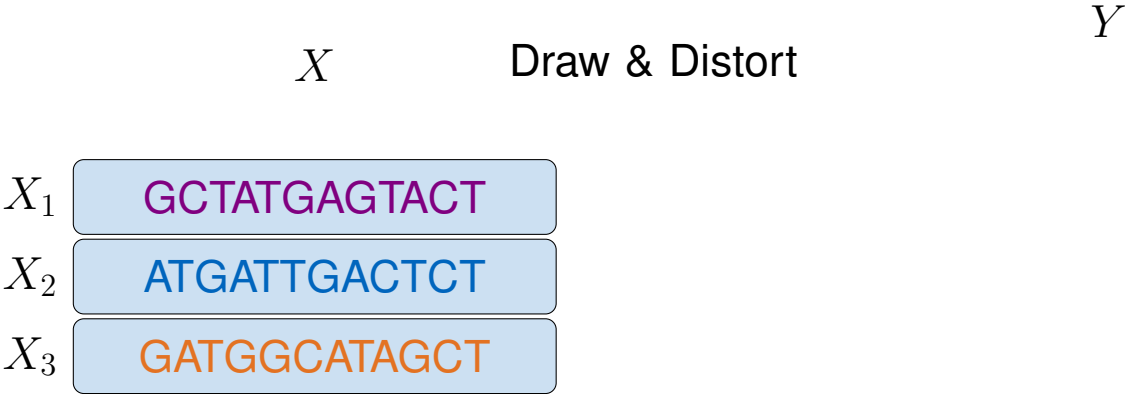
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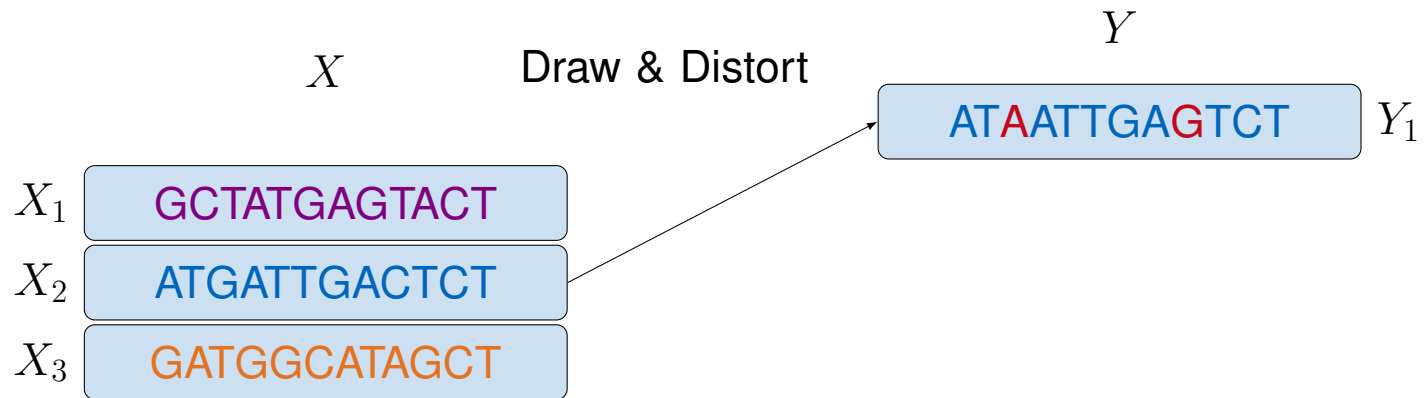
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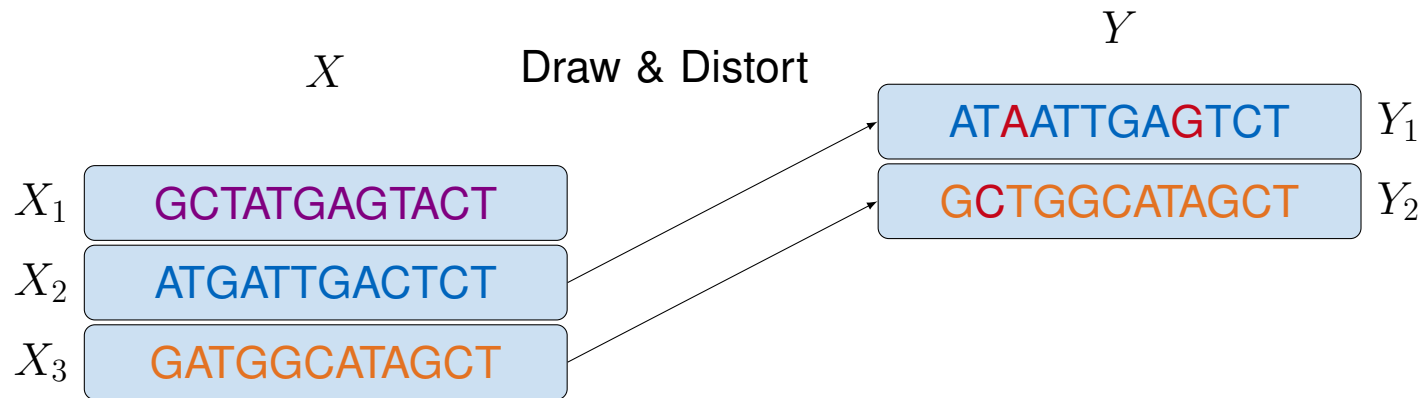
Channel Model



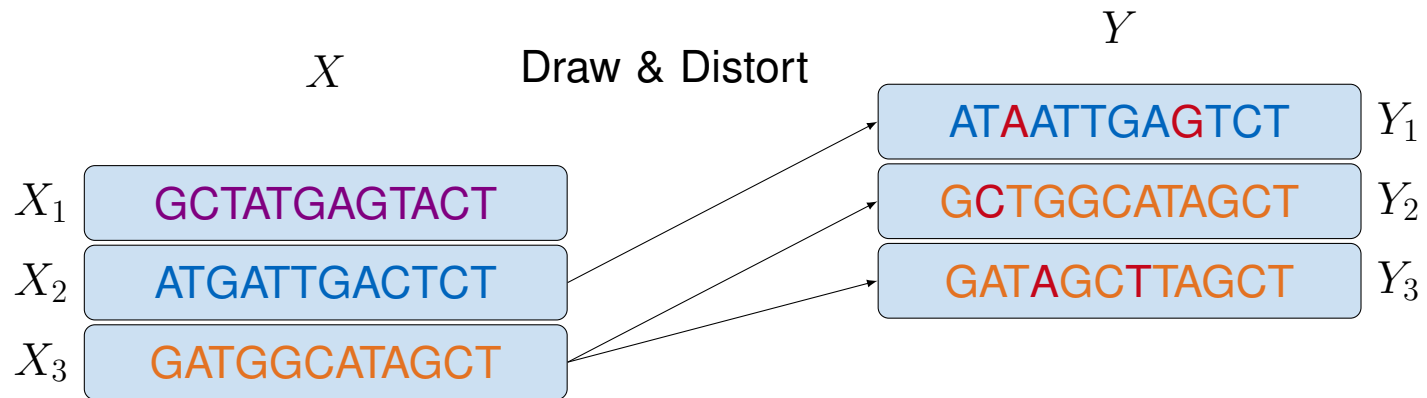
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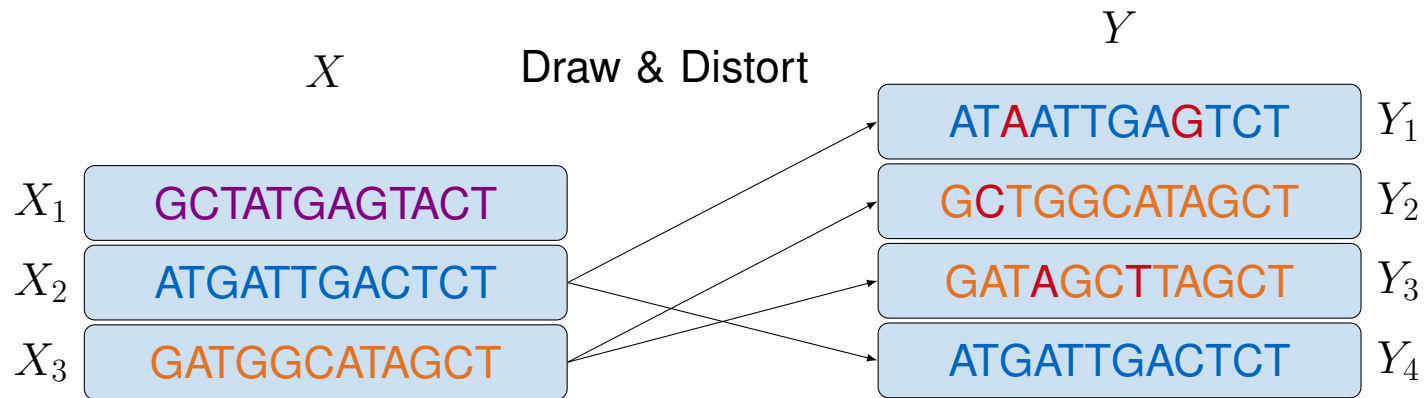
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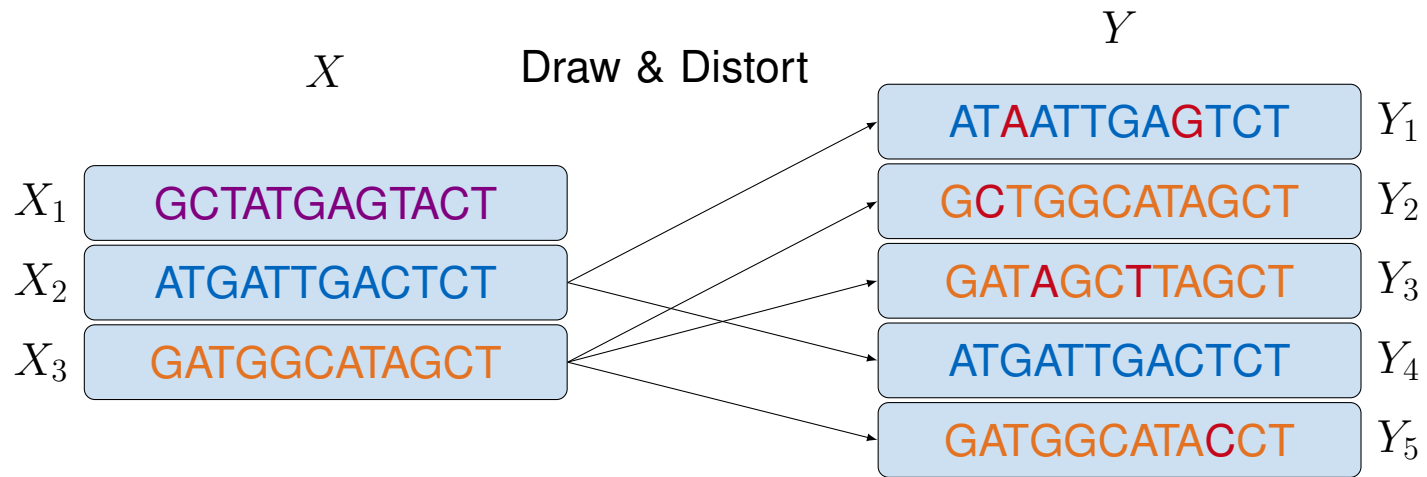
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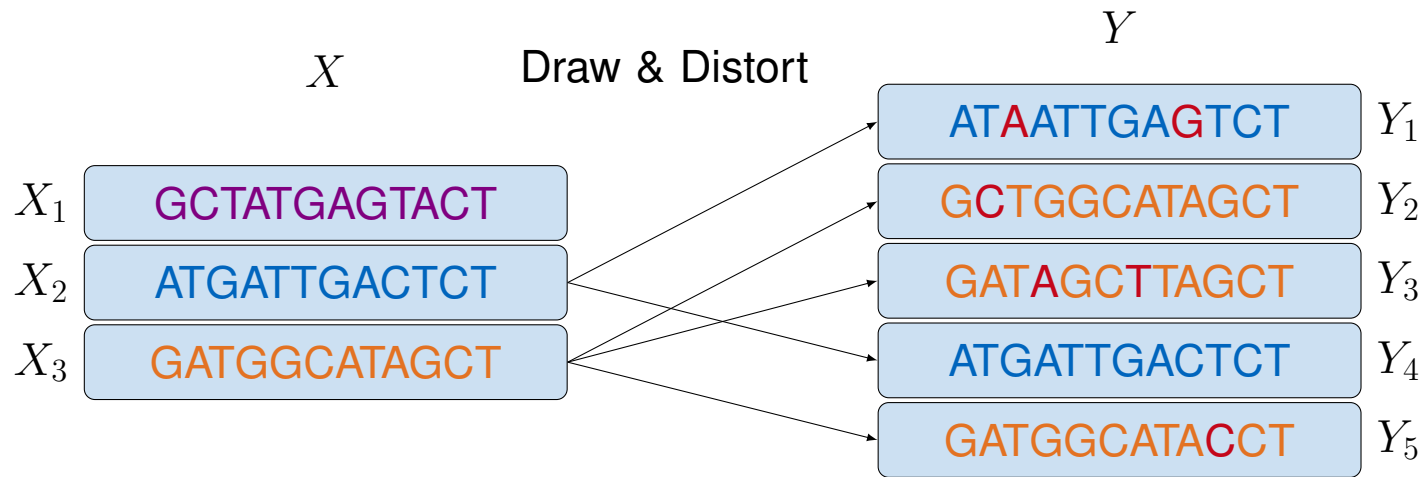
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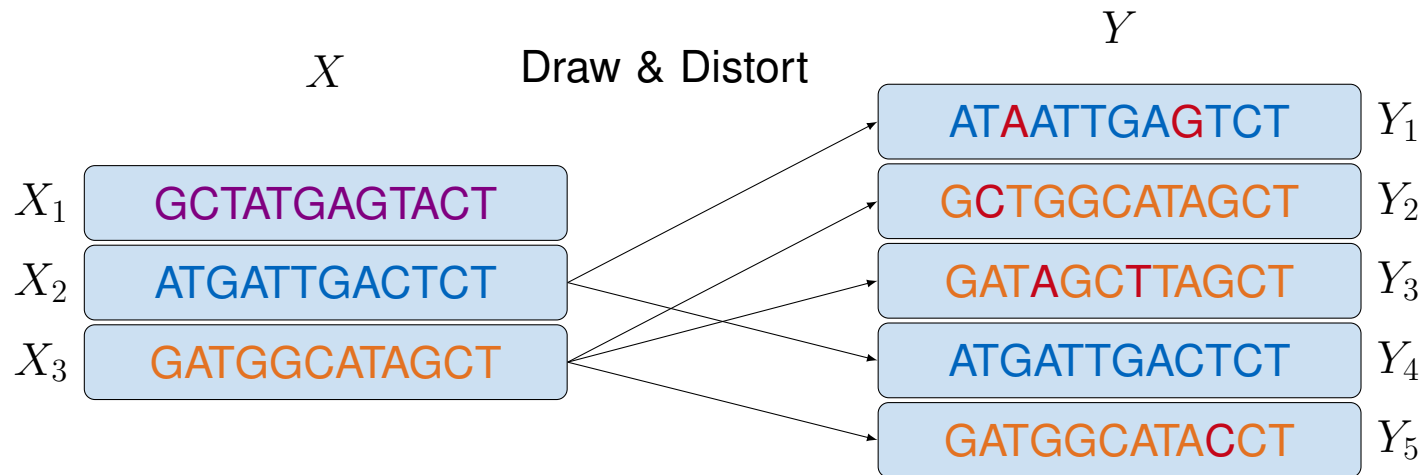
Channel Model



$$Y_j = X_{I_j} + E_j$$

- I_j : i.i.d. uniform random draws
- E_j : random error vectors (error probability p)

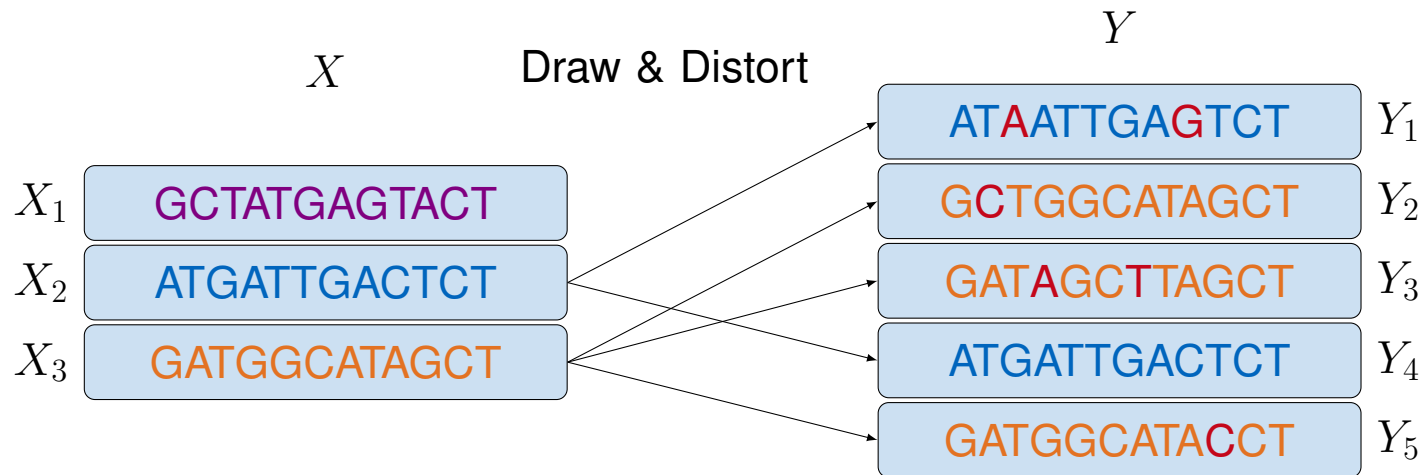
Channel Model



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- In this work: Quaternary sequences ($\mathbb{Z}_4 = \{A, C, G, T\}$)

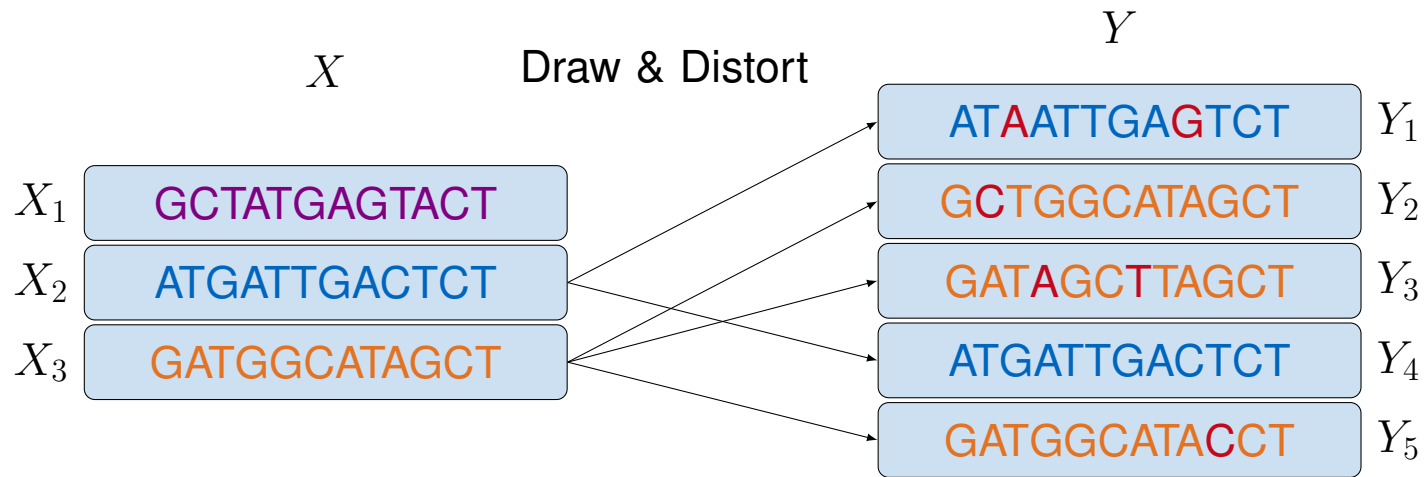
Channel Model



Channel Input:

- M Sequences, each of length L
- $X = (X_1, \dots, X_M)$

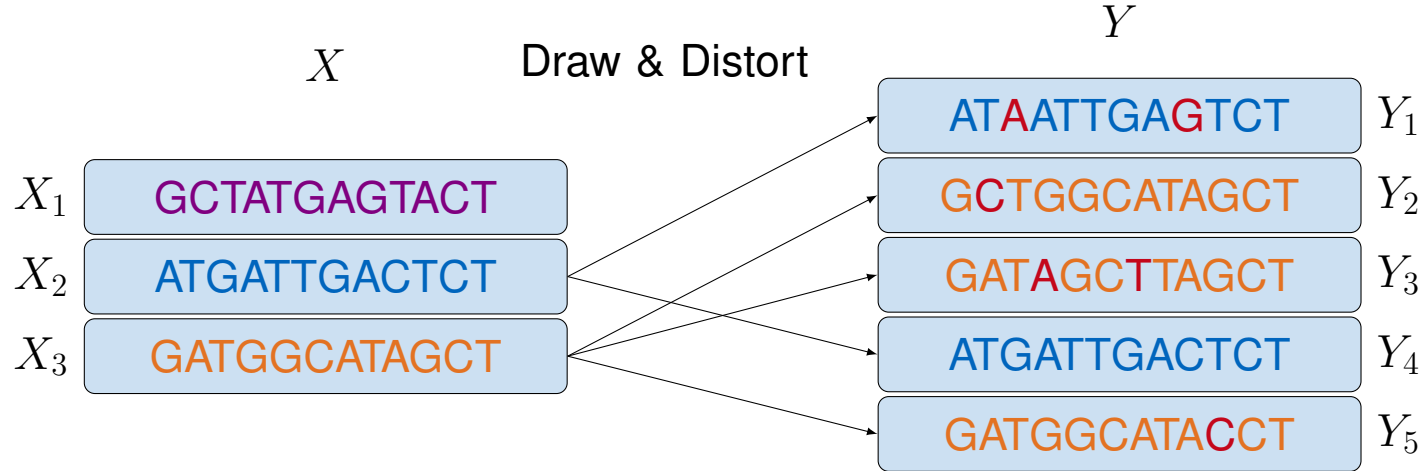
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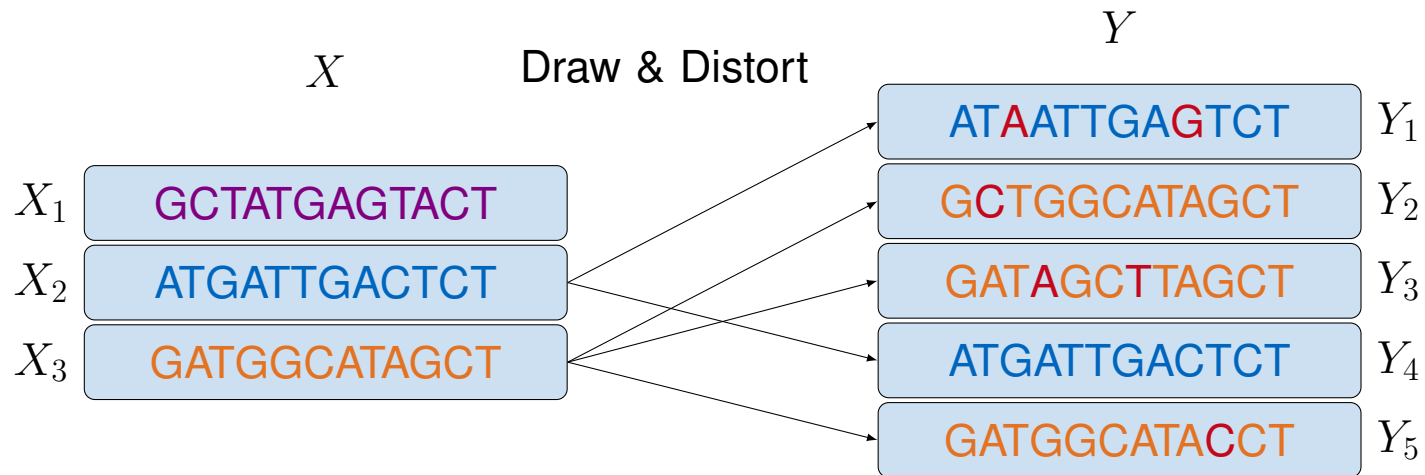
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Channel Model - Codes and Information Rates

Communication System:

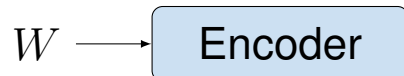
Message

W

Channel Model - Codes and Information Rates

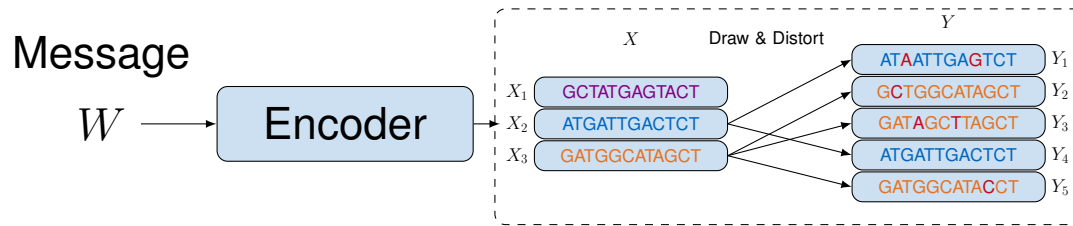
Communication System:

Message



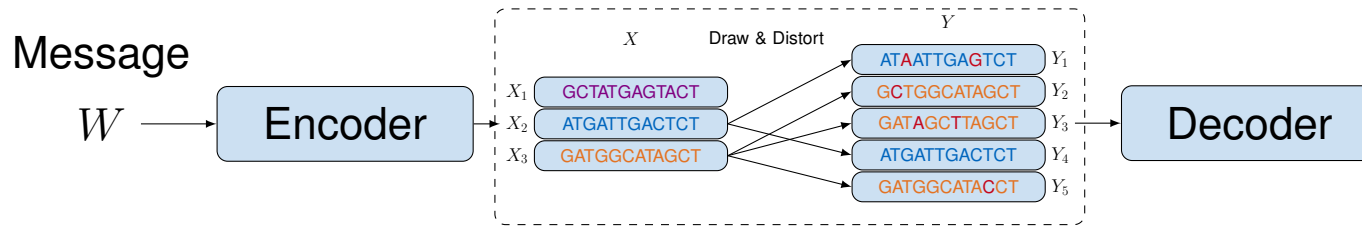
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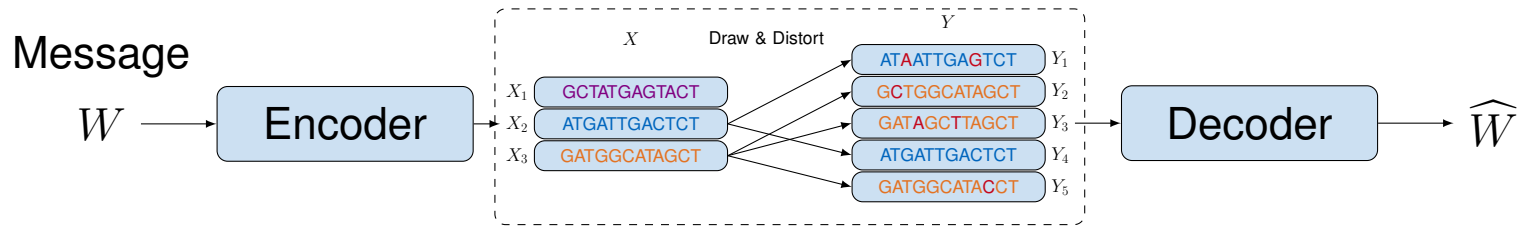
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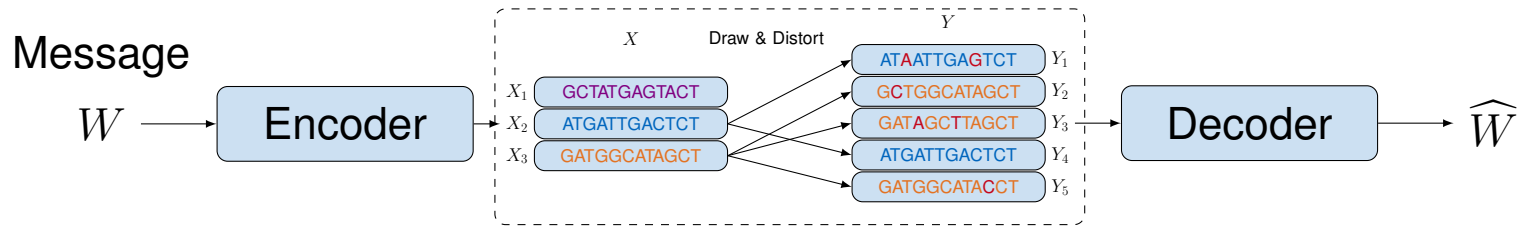
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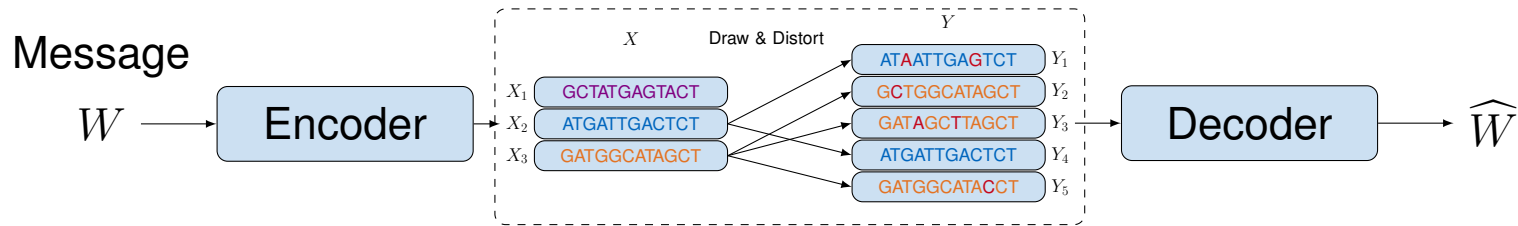


Code:

- $\mathcal{C} = \{X(1), \dots, X(4^{MLR})\} \subset \mathbb{Z}_4^{M \times L}$
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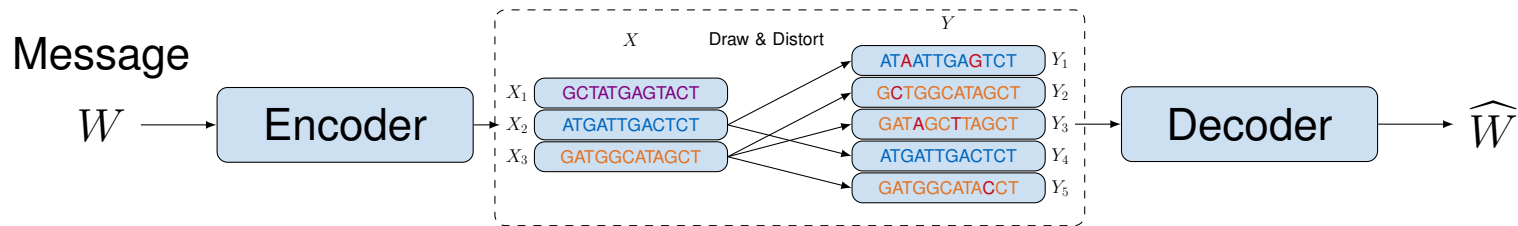
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- Error prob. $P(\text{Err}) = P(\text{dec}(Y) \neq X)$

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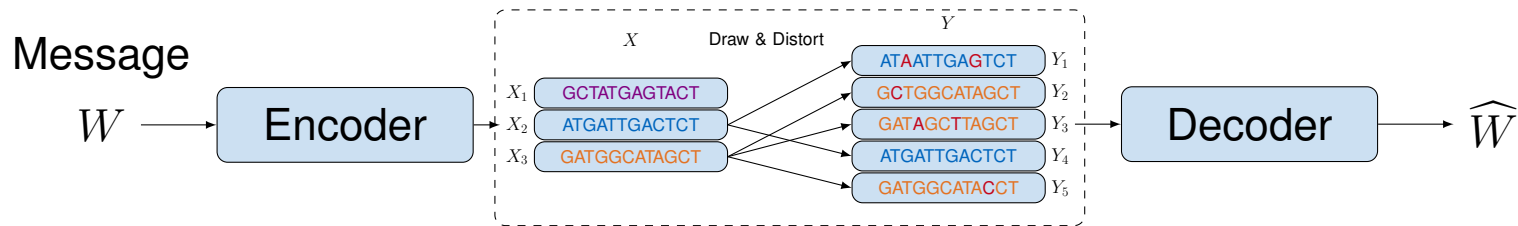
Channel Capacity

Achievable rates

Code rate R is achievable, if there exists a code \mathcal{C} of rate R with $P(\text{Err}) \rightarrow 0$, as $ML \rightarrow \infty$

Channel Model - Codes and Information Rates

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Achievable rates

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- **Capacity:** Supremum of achievable rates

Related Work

[Mitzenmacher, "On the Theory and Practice of Data Recovery with Multiple Versions," 2006]

- Capacity of binomial/multi-draw channel

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- Generalization to asymmetric channels
- Computation of error probabilities

Preliminaries - Channel Model Revisited

Alternative Channel Formulation

X

X_1

GCTATGAGTACT

X_2

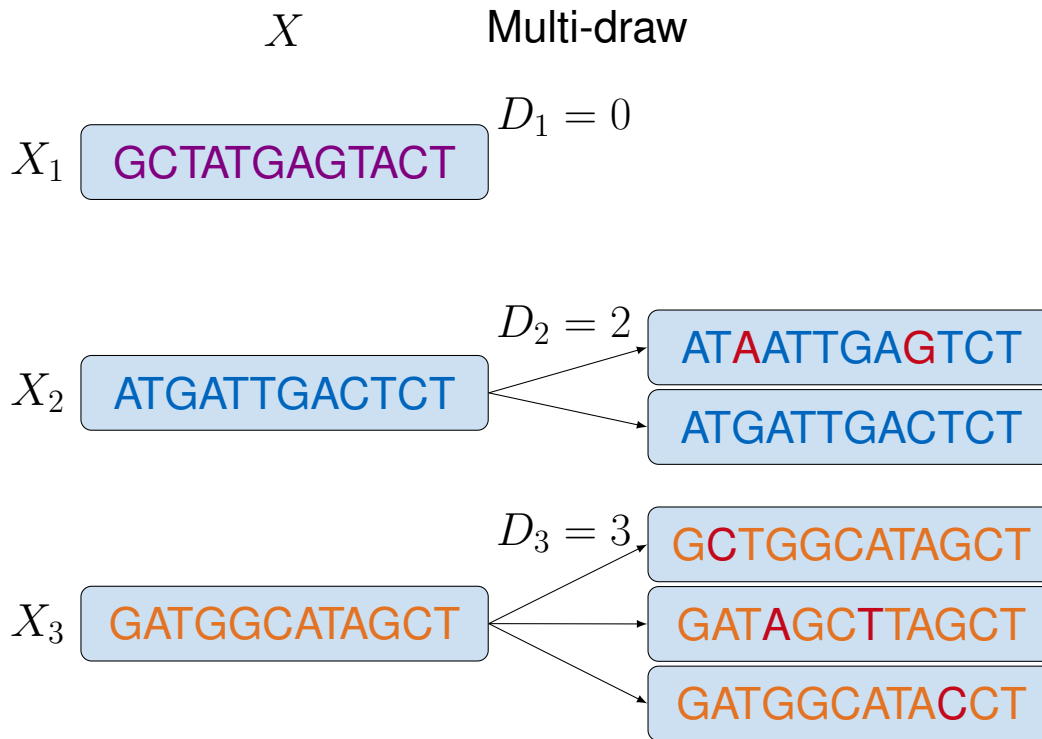
ATGATTGACTCT

X_3

GATGGCATAGCT

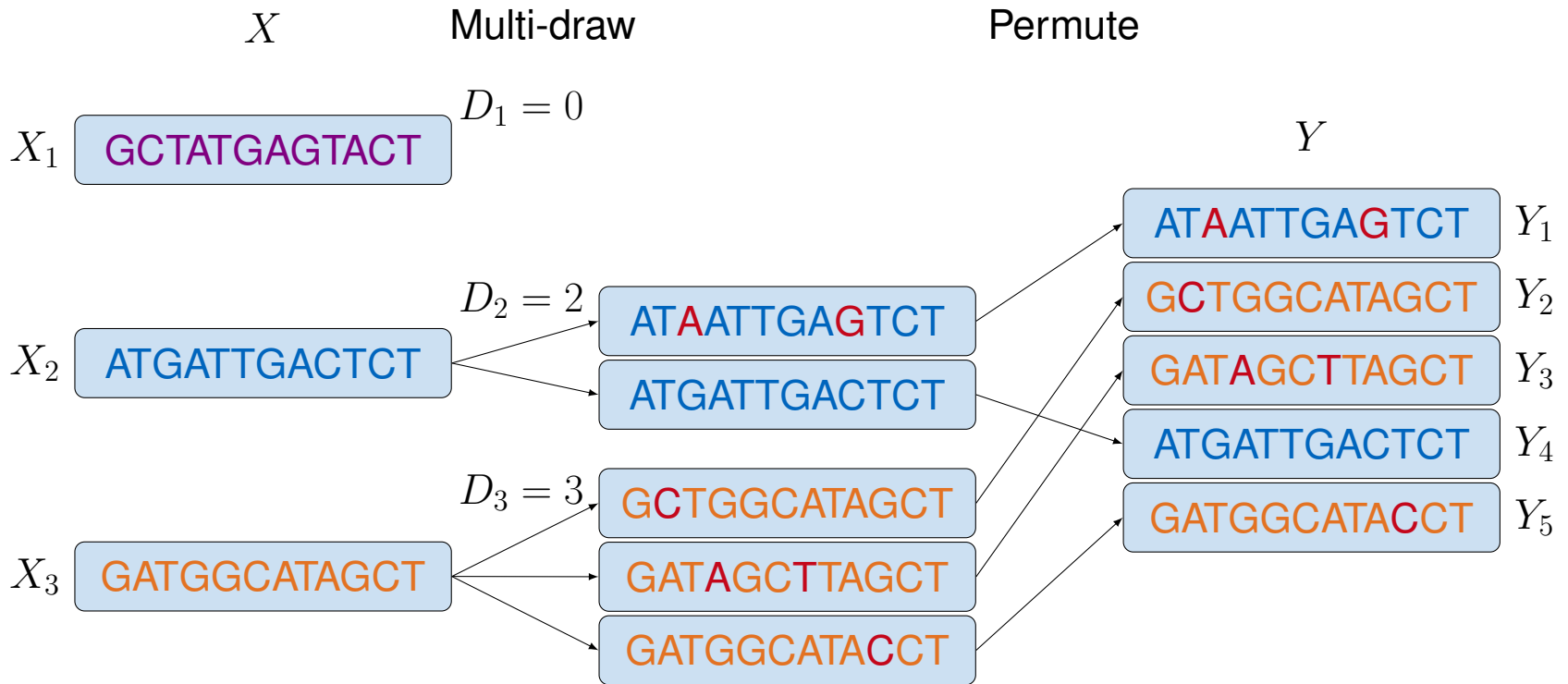
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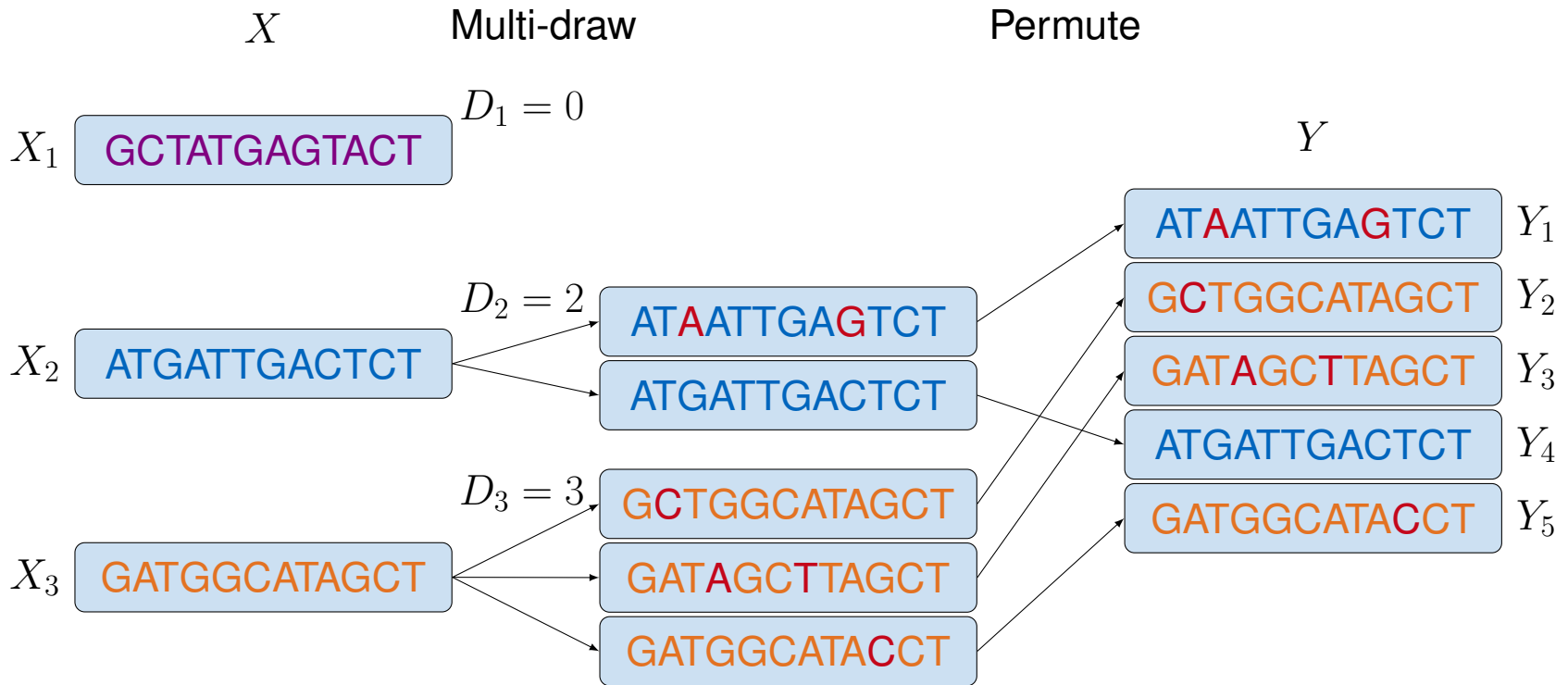
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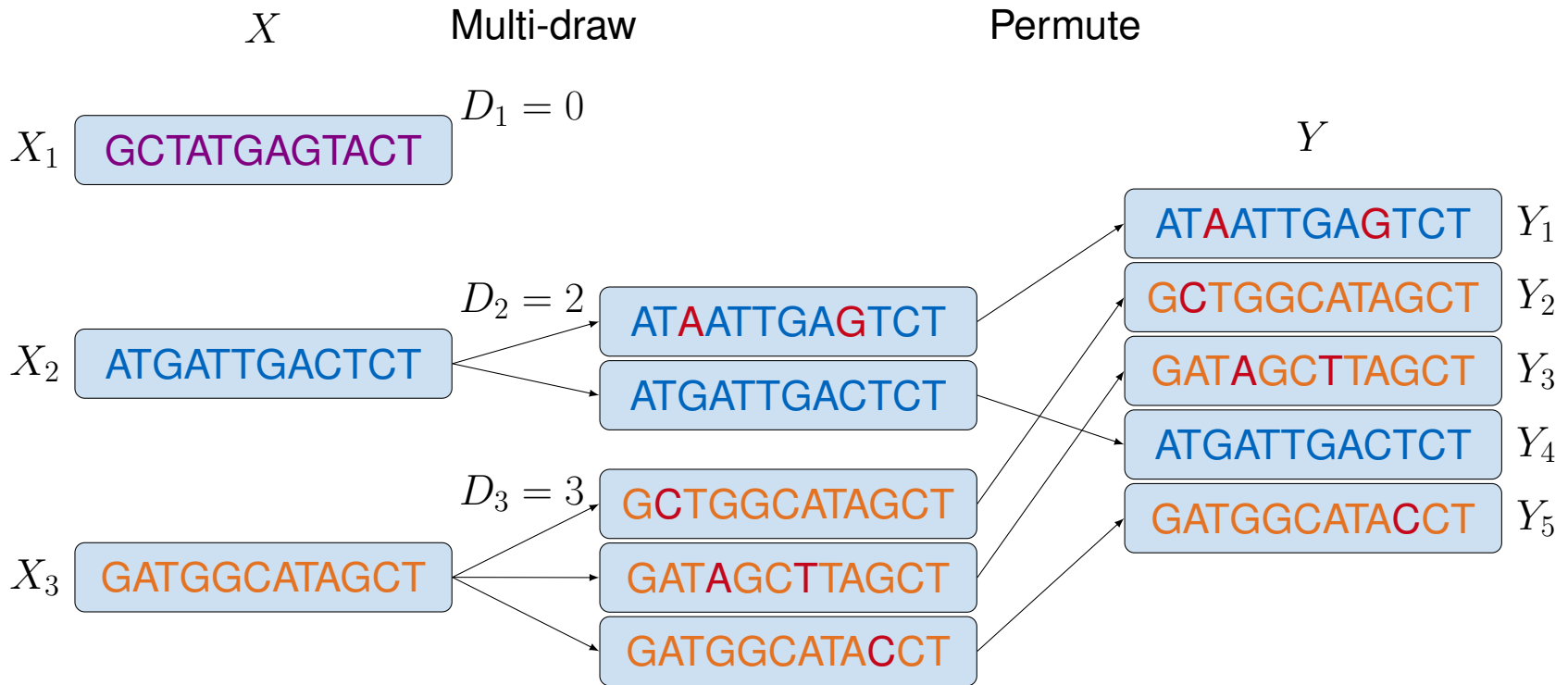


Challenges

- Draws of the multi-draw channels are random

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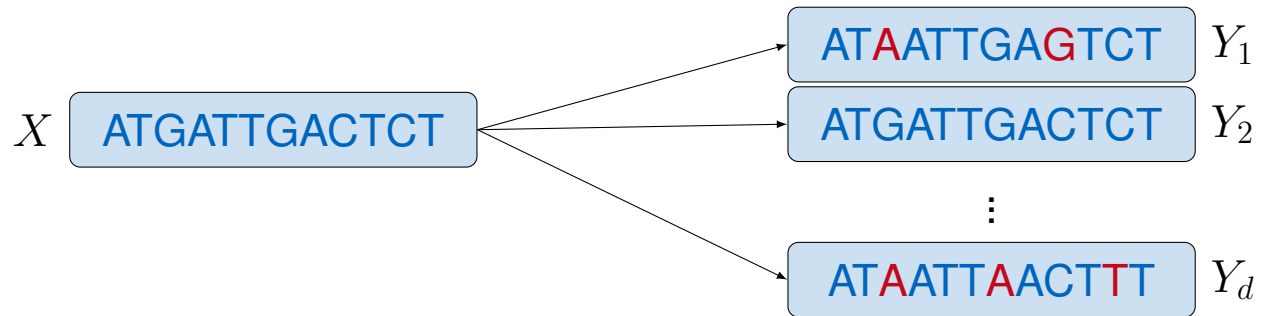
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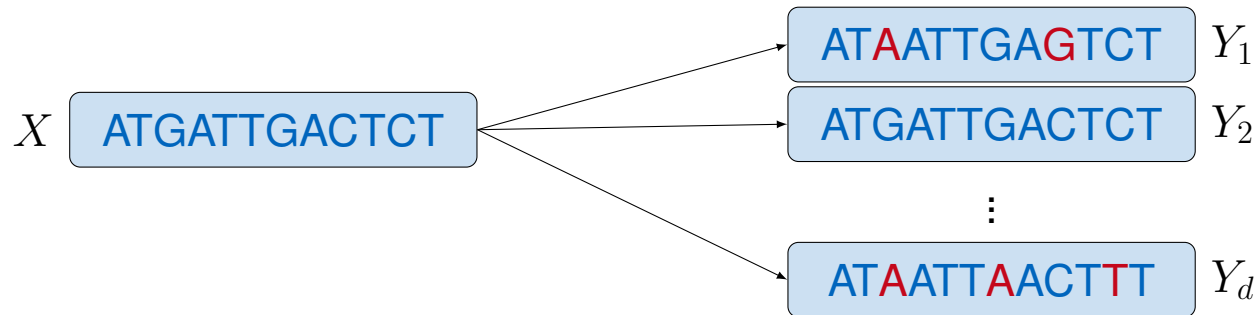
Challenges

- Draws of the multi-draw channels are random
- Permutation of the sequences

Preliminaries - Multi-draw Channel [Mitzenmacher, 2006]

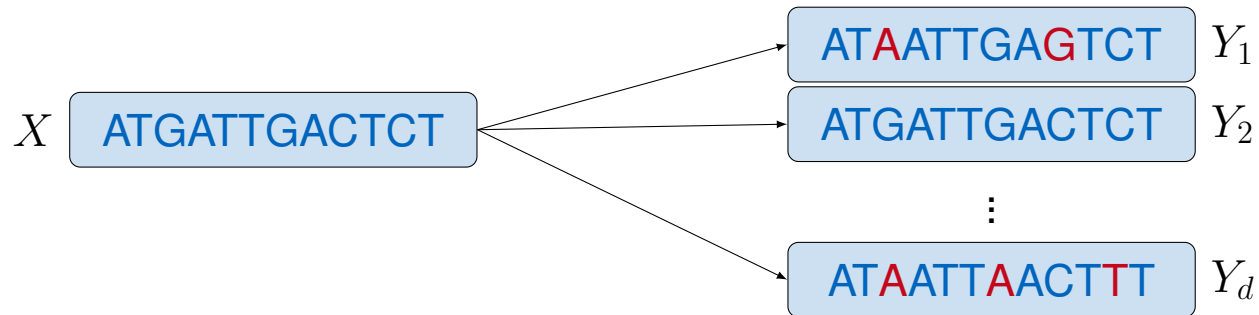


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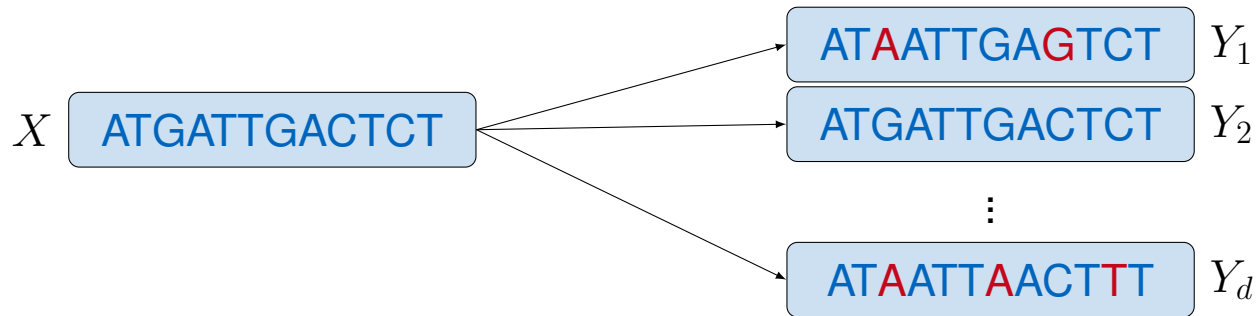
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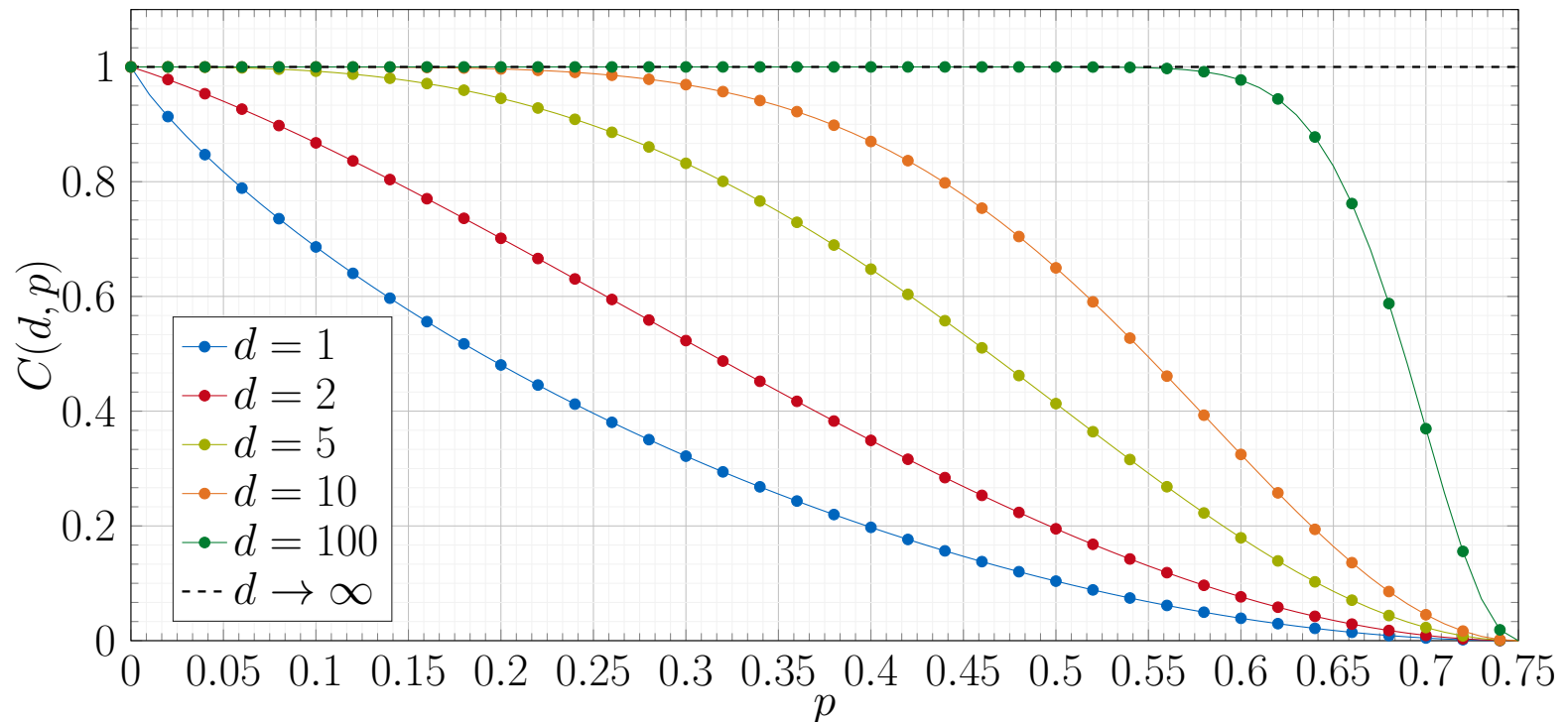


- Input: X
- Output: d output sequences Y_1, \dots, Y_d
- q -ary symmetric channels: $Y_i = X + E_i$

Preliminaries - Multi-draw Channel [Mitzenmacher, 2006]



- Capacity (d draws, error probability p)



Channel Capacity

Draw Distribution

- Recall:
 - ▶ D_i : Number of draws of sequence i

Channel Capacity

Draw Distribution

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Channel Capacity

Theorem: Channel Capacity

Given $2\beta < 1 - H_4(2p)$, the capacity is

$$C(c, \beta, p) = \sum_{d=0}^{\infty} \text{Poi}(c, d) C_{\text{Mul}}(d, p) - \beta(1 - e^{-c})$$

Channel Capacity - Parameter Range

Parameter Range

$$2\beta < 1 - H_4(2p)$$

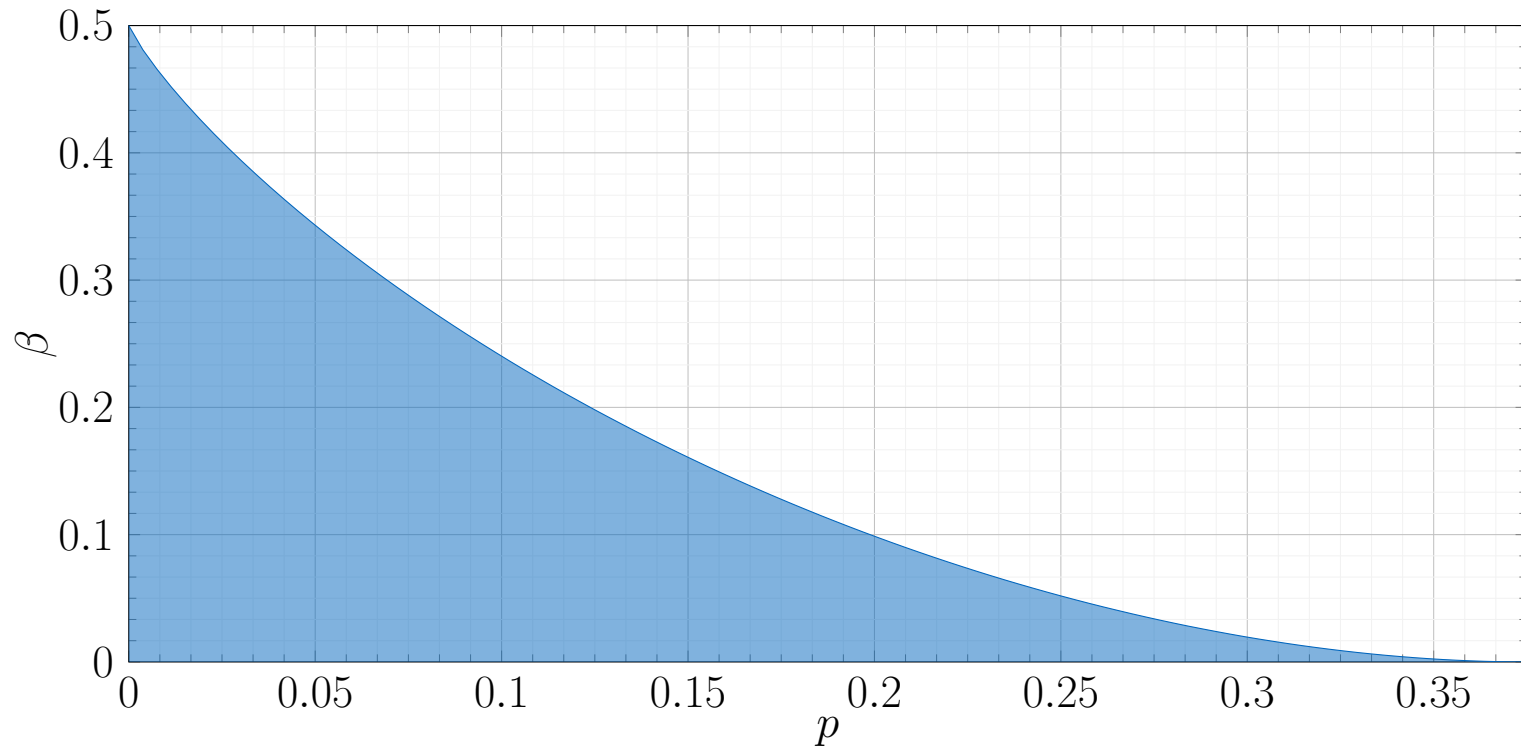
- Entropy function $H_4(p) = -(1 - p) \log_4(1 - p) - p \log_4\left(\frac{p}{3}\right)$

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Channel Capacity - Discussion

Capacity

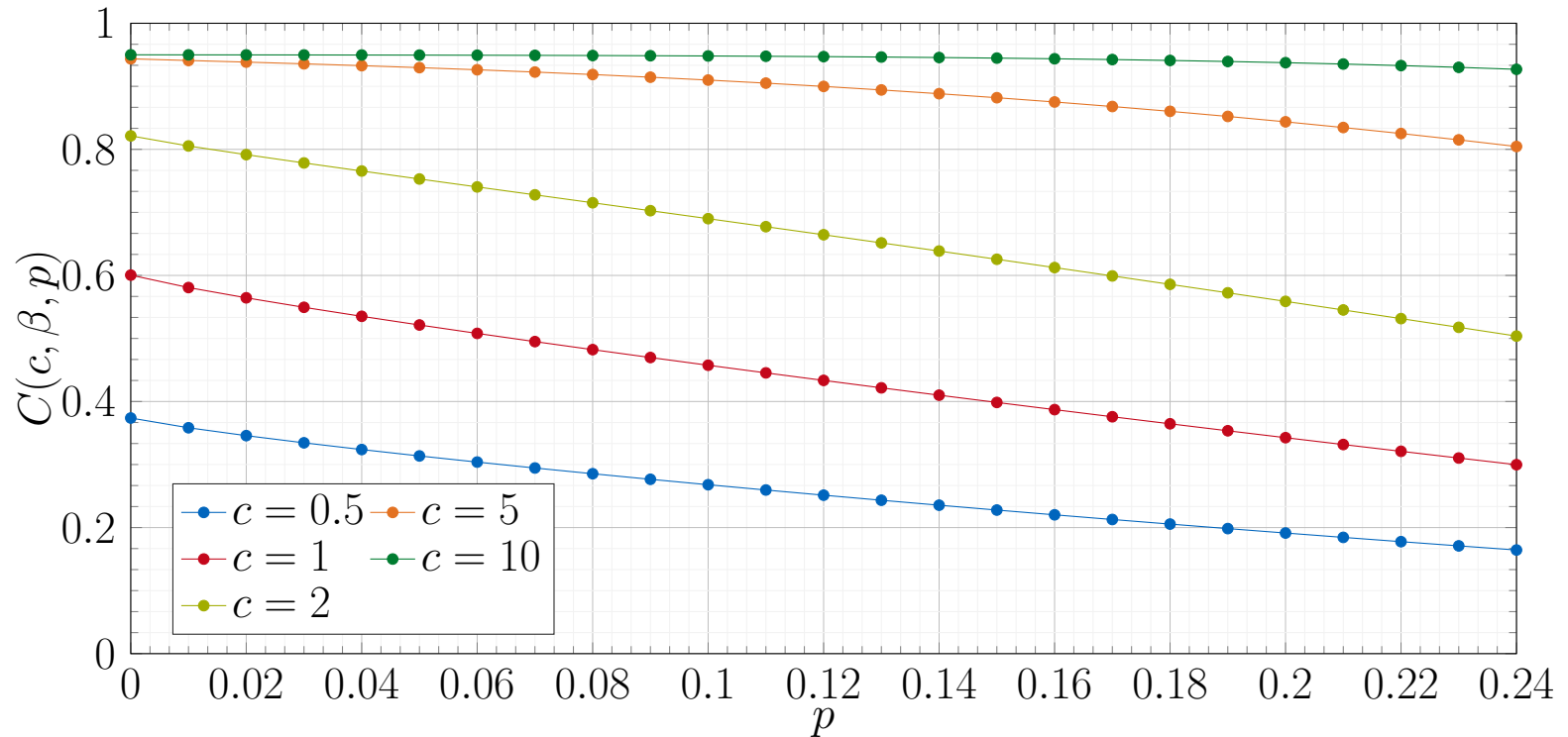
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$$\beta = 1/20$$



Channel Capacity - Discussion

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Channel Capacity - Storage and Recovery Rate Tradeoff

Storage and Recovery Rate Tradeoff

- Storage rate: $R_s = \log_2 |\mathcal{C}| / ML$

Channel Capacity - Storage and Recovery Rate Tradeoff

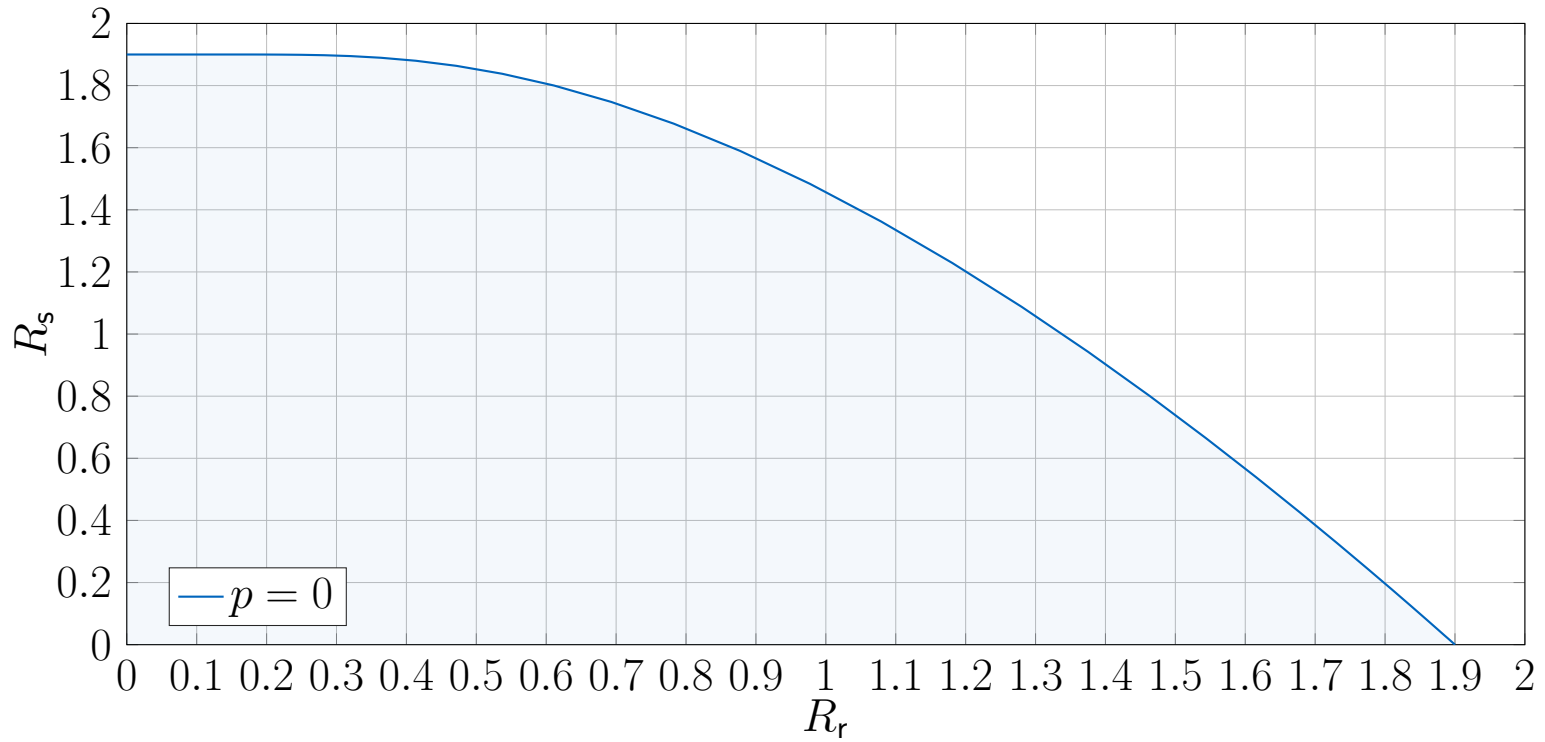
Storage and Recovery Rate Tradeoff

- Storage rate: $R_s = \log_2 |\mathcal{C}| / ML$
- Recovery rate: $R_r = \log_2 |\mathcal{C}| / NL$

Channel Capacity - Storage and Recovery Rate Tradeoff

Storage and Recovery Rate Tradeoff

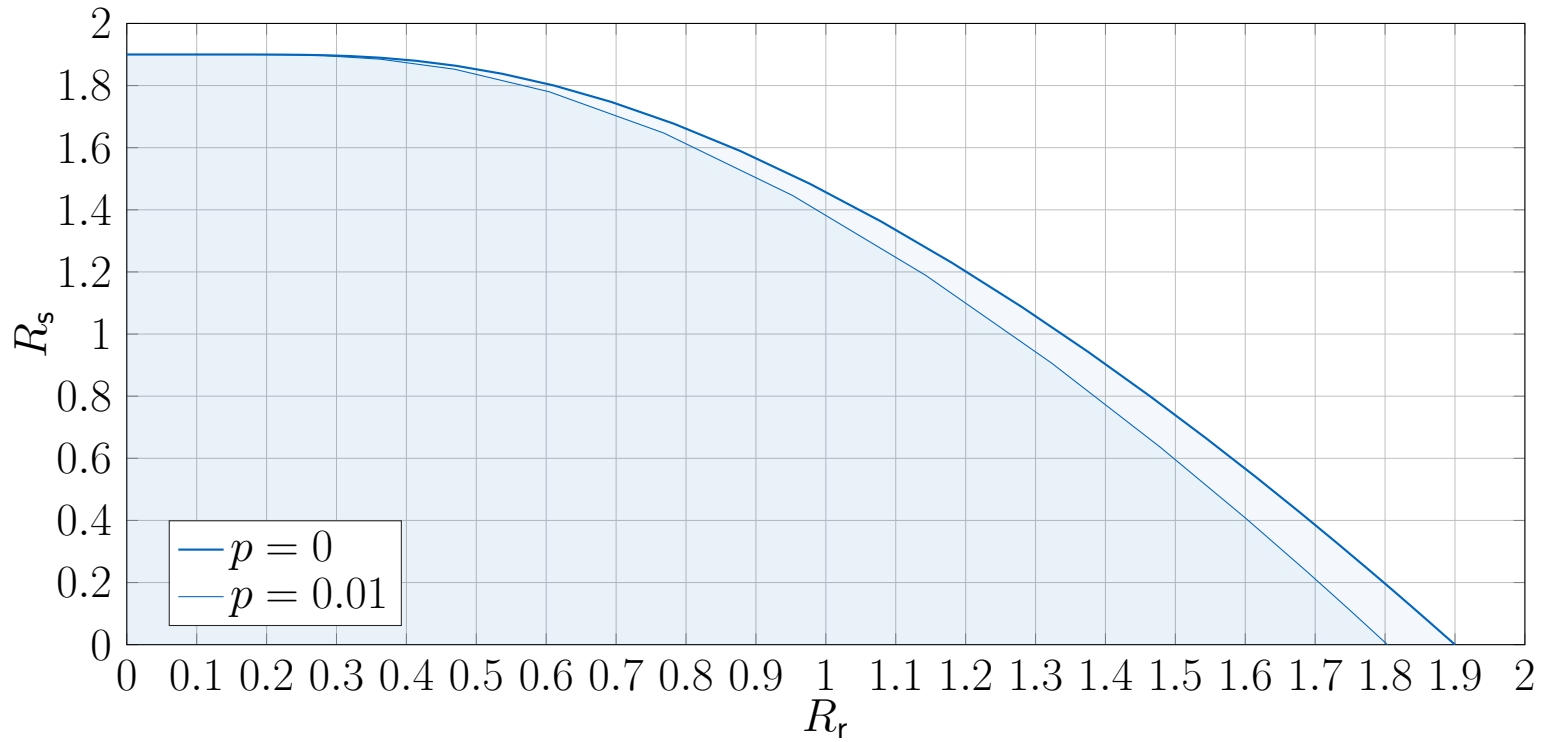
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Channel Capacity - Storage and Recovery Rate Tradeoff

Storage and Recovery Rate Tradeoff

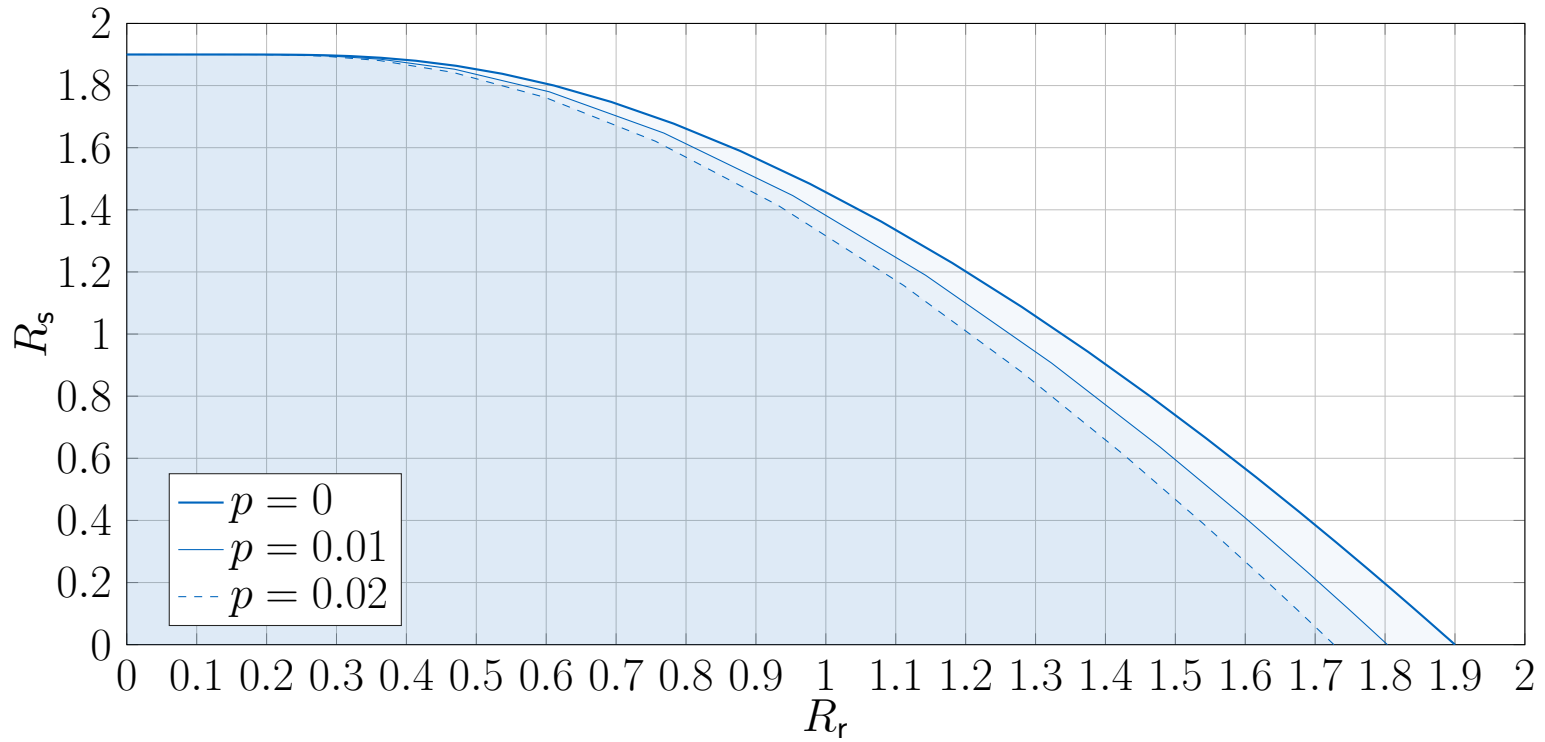
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Channel Capacity - Storage and Recovery Rate Tradeoff

Storage and Recovery Rate Tradeoff

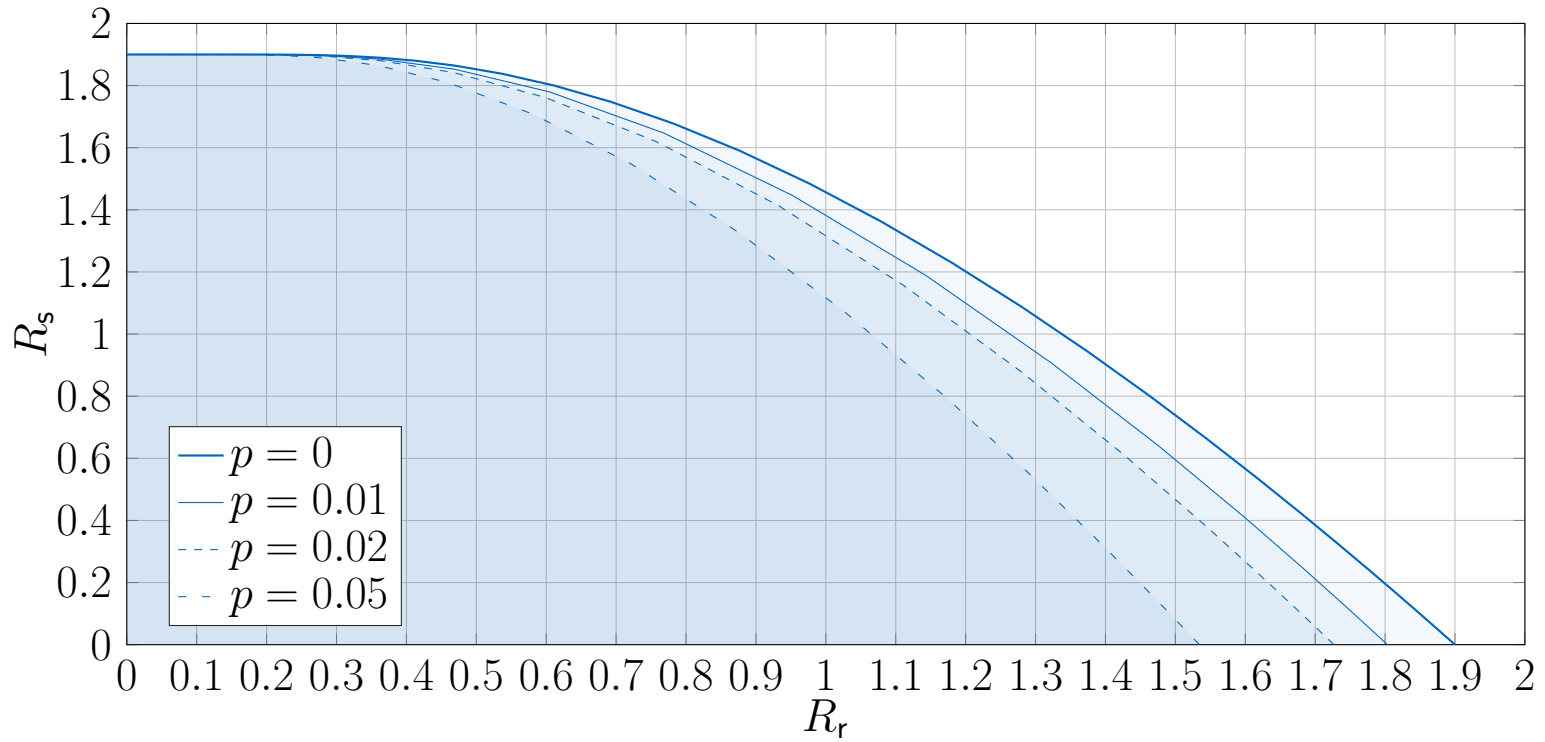
- Storage rate: $R_s = \log_2 |\mathcal{C}| / ML$
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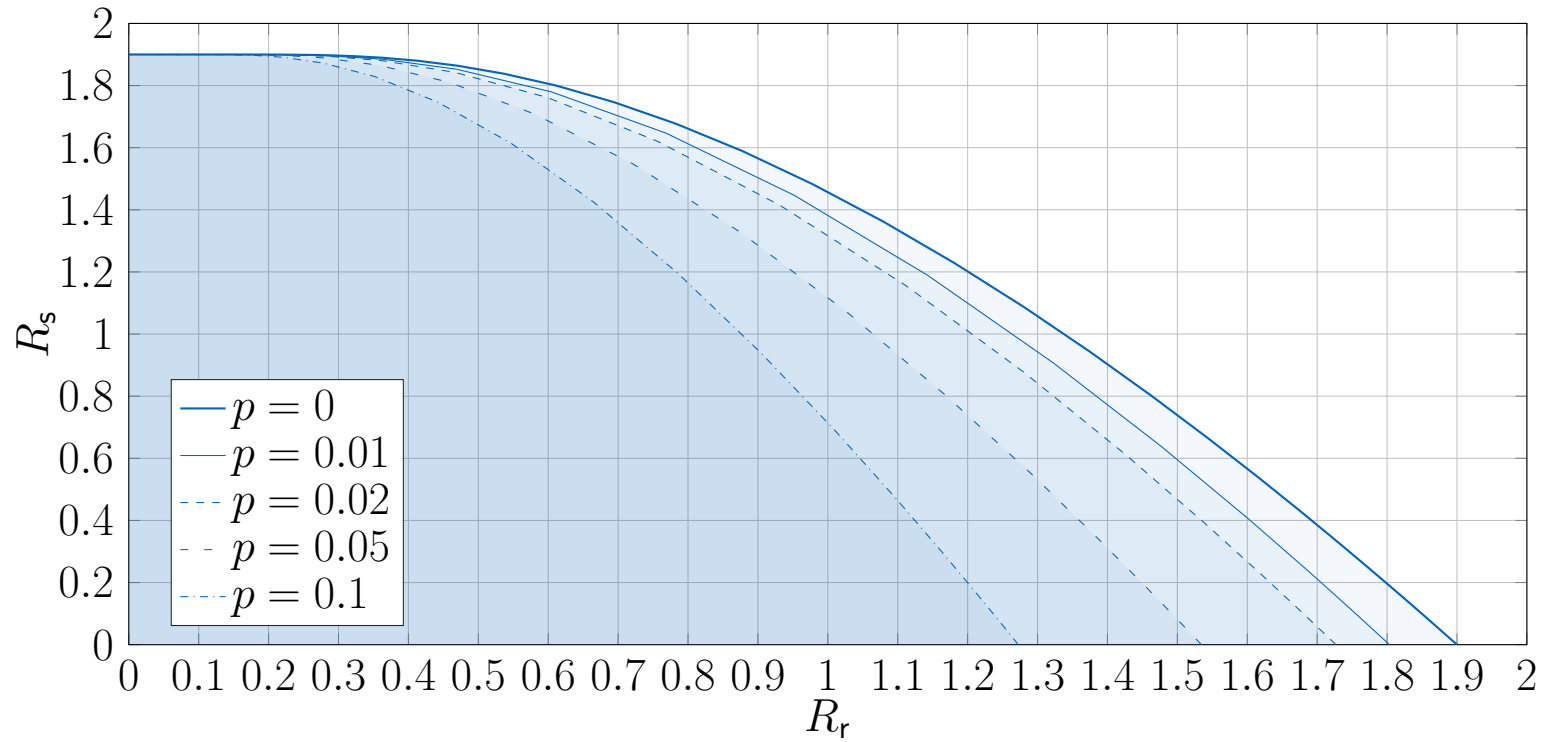
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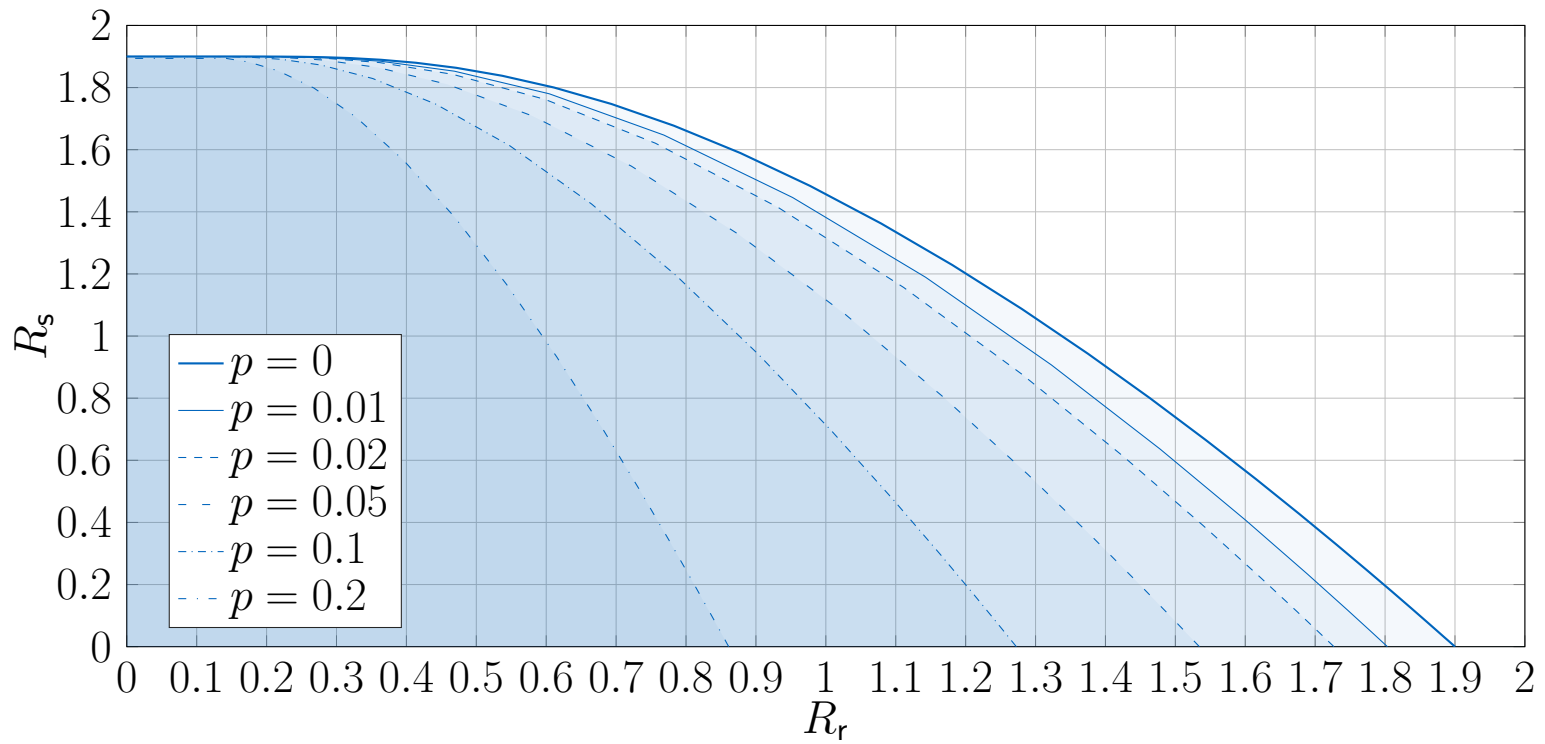
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- Challenges for efficiently encodable/decodable schemes

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Thank you!