

# **Erasure Probabilities and Error Correction Assignment**

**Submitter:** Ron Bartov

# Background and Introduction

This document provides answers and simulations analysis for the post-interview assignment, which focuses on implementing coding functions and understanding theoretical concepts related to bit-channel erasure probabilities, encoding, channeling, decoding, and frame error rate (FER) analysis.

The assignment is divided into two parts:

- **Part 1:** Focuses on calculating erasure probabilities for bit-channels and analyzing how parameters influence these probabilities.
- **Part 2:** Explores encoding, channel operations, decoding, and their associated error probabilities.

The following sections provide detailed answers, function explanations, and simulations analysis.

## Preliminary Knowledge

This section includes all the diagrams and equations from the assignment, serving as the basis for later calculations and analysis.

### U-Transform

A transformation from vector  $x_n$  to vector  $u_n$

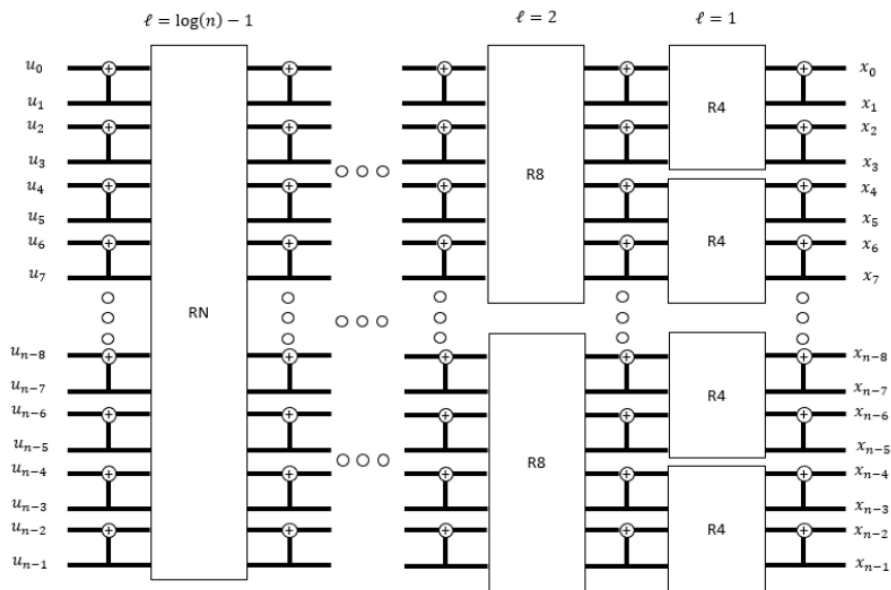


Figure 1: U-transform

### Permutation Block Basic Unit for Layer $l$

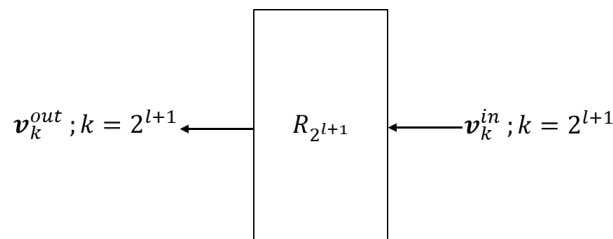


Figure 2: Permutation block basic unit for layer  $l$

This permutation block performs the following task:

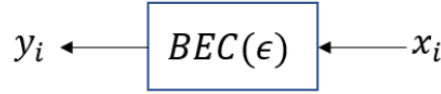
For  $j = 0 \dots 2^{\ell+1} - 1$

- If index  $j < 2^\ell$ :
  - $v_{2j}^\ell \leftarrow v_j^\ell$
- Else
  - $v_{2(j-2^\ell)+1}^\ell \leftarrow v_j^\ell$

Figure 3: Permutation algorithm for permutation block basic unit

## BEC( $\epsilon$ ) Channel

A communication channel model where each transmitted bit has a probability  $\epsilon$  of being erased



$$y_i = \begin{cases} x_i & \text{w.p. } 1 - \epsilon \\ E & \text{w.p. } \epsilon \end{cases}$$

Figure 4: BEC channel

## Bit Channel Decoder (BCD) Basic Unit

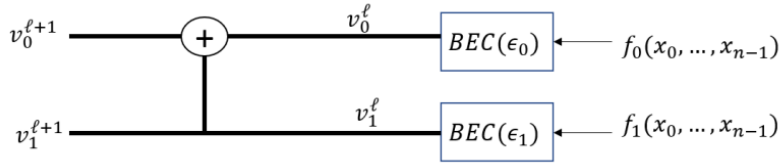


Figure 5: Bit channel decoder basic unit

Satisfies the following effective channel equations for  $v_j^{l+1}$ :

For  $j = 0$ , assuming  $\mathbf{x}$  is given:

$$\epsilon_0^{l+1} = p(v_0^{l+1} = E | \mathbf{x}) = 1 - (1 - \epsilon_0^\ell)(1 - \epsilon_1^\ell)$$

Equation 1: Erasure probability for upper output of BCD basic unit

For  $j = 1$ , assuming  $v_0^{l+1}$  and  $\mathbf{x}$  are given:

$$\epsilon_1^{l+1} = p(v_1^{l+1} = E | \mathbf{x}, v_0^{l+1} \text{ given}) = \epsilon_0^\ell \epsilon_1^\ell$$

Equation 2: Erasure probability for lower output of BCD basic unit

For  $j > 1$ , assuming all  $v_{j'}^{l+1}, j' < j$  and  $\mathbf{x}$  are given:

$$\epsilon_j^{l+1} = p(v_j^{l+1} = E | \mathbf{x}, v_0^{l+1}, \dots, v_{j-1}^{l+1} \text{ given})$$

Equation 3: Definition of erasure probability for array elements, that are the outputs of concatenate BCD units, for index greater than one

## Encoder-Channel-Decoder (ECD) Model for Input Vector of Length 4

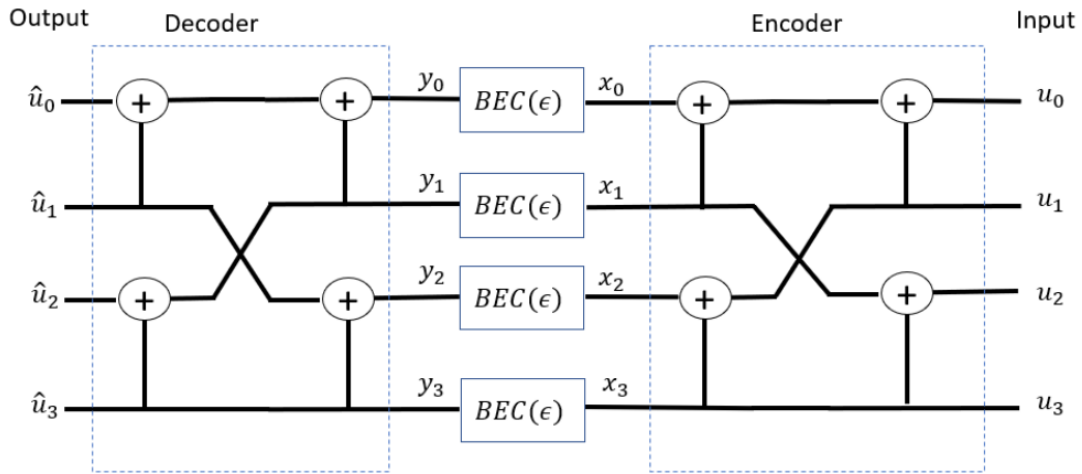


Figure 6: Encoder-Channel-Decoder Model for part 2

Satisfies the following output vector equations:

$$\hat{u}_0 = y_0 + y_1 + y_2 + y_3$$

Equation 4: ECD output 0

$$\hat{u}_1 | \hat{u}_0 = y_2 + y_3 \text{ or } \hat{u}_0 + y_0 + y_1$$

Equation 5: ECD output 1

$$\hat{u}_2 | \hat{u}_1, \hat{u}_0 = y_1 + y_3 \text{ or } (\hat{u}_0 + \hat{u}_1) + y_0 + y_3 \text{ or } \hat{u}_0 + y_0 + y_2 \text{ or } y_1 + y_2 + \hat{u}_1$$

Equation 6: ECD output 2

$$\hat{u}_3 | \hat{u}_2, \hat{u}_1, \hat{u}_0 = y_3 \text{ or } \hat{u}_1 + y_2 \text{ or } \hat{u}_2 + y_1 \text{ or } \hat{u}_0 + \hat{u}_1 + y_0 + \hat{u}_2$$

Equation 7: ECD output 3

# Part 1: Bit-Channel Erasure Probabilities and Transform Analysis

This part investigates the behavior of bit-channels in a  $BEC(\epsilon)$ . It examines how parameters such as the erasure probability  $\epsilon$  and the size of the transform  $n$  affect the reliability of bit-channels. Insights are drawn through computations, sorting, and plotting of erasure probabilities.

## Question 1: Write a function that computes an array of erasure probabilities

### Function Signature:

```
def erasure_probability_recursive_u_transform(n: int, epsilon: int) -> np.ndarray:
```

### Functions Used:

- **effective\_bit\_channels\_in\_bcd:**  
Implements the transition of an array through a series of blocks of “[Bit Channel Decoder Basic Unit](#)”.
- **permutation\_block:**  
Implements the “[Permutation Block](#)”.
- **verify\_power\_of\_two:**  
Verify that  $n$  is a power of two (used for validation through the simulations)

### Algorithm:

- **Base Case:**  
If  $n = 2$ :
  - Initialize the input erasure probabilities as  $[\epsilon, \epsilon]$ .
  - Apply the “[Bit Channel Decoder Basic Unit](#)” to compute the erasure probabilities.
  - Return the computed probabilities.
- **Recursive Case:**  
If  $n > 2$ :
  - Compute the layer index by  $l = \log(n) - 1$ .
  - Recursively calculate the erasure probabilities for two vectors with size  $\frac{n}{2}$ .
  - Concatenate the computed probabilities for the half-size blocks to create the input probabilities vector for the current block size.
  - Apply the permutation operation based on the current layer  $l$  to reorder the input probabilities.
  - Apply the “effective\_bit\_channels\_in\_bcd” function on the reordered probabilities.
  - Return the computed probabilities.

**Example for  $N = 8$ :**

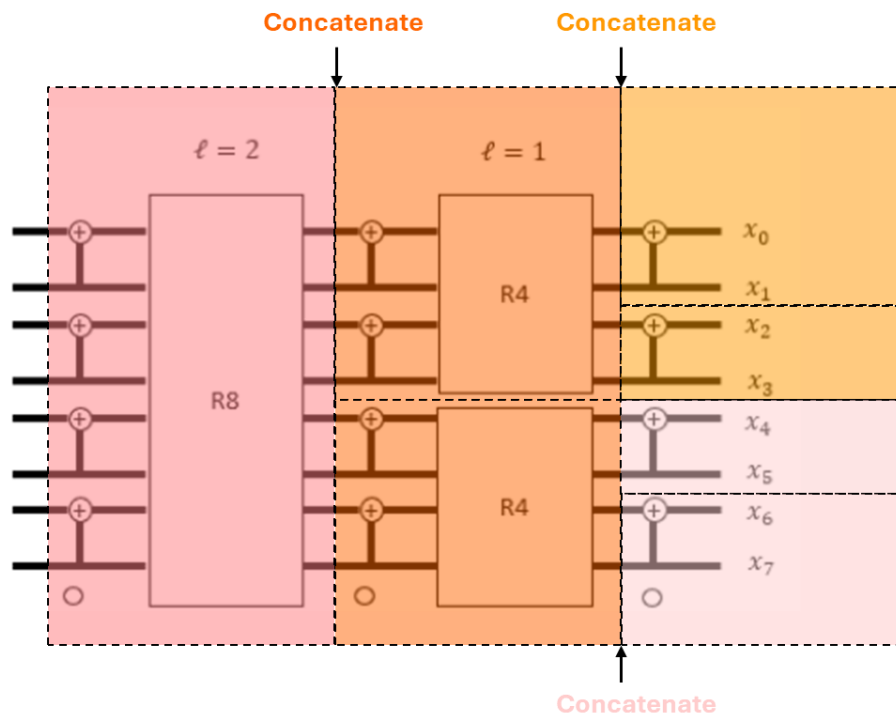


Figure 7: Recursive U-transform example for  $N=8$

**Question 2: Simulate erasure probabilities for different input lengths**

**Simulation Results:**

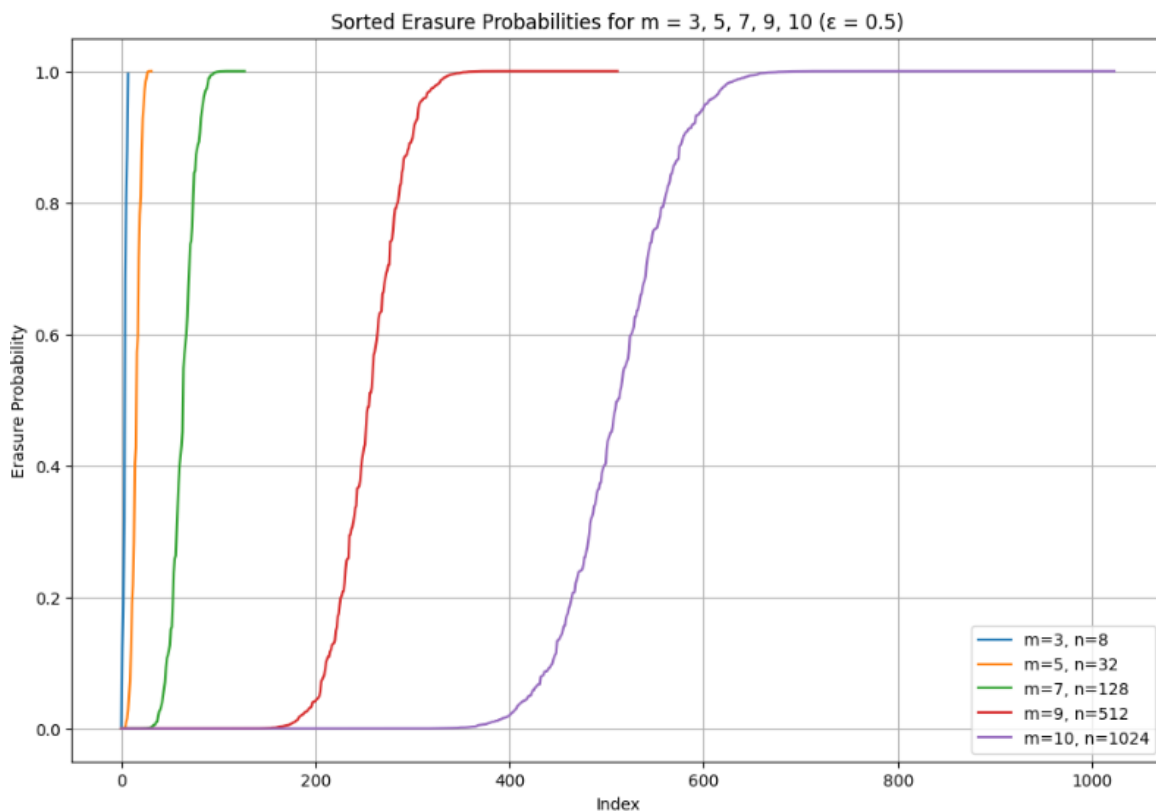


Figure 8: Erasure probabilities simulation for different input lengths and for  $\epsilon = 0.5$

### Simulation Analysis:

- From the plot, we observe that for a given  $n$  the erasure probabilities span the entire range between 0 and 1. This indicates that some output indices exhibit low erasure probabilities, while others have high erasure probabilities.
- The graph's shape (based on the selected axes) resembles a shifted sigmoid curve. Notably, the number of output indices with erasure probabilities less than 0.5 is approximately equal to those with erasure probabilities greater than 0.5.
- As  $n$  increases, the graph retains its overall shape, and the erasure probability values are still distributed evenly around 0.5.
- The [U-Transform](#) is composed of multiple layers, where each layer concatenates the outputs of [Bit Channel Decoder Basic Units](#) and passes them through [Permutation Block Basic Units](#). Within a given layer, based on the equations provided for the [Bit Channel Decoder Basic Unit](#), an input erasure probability vector of length 2 yields two distinct erasure probability outputs:  $\epsilon^2$  and  $1 - (1 - \epsilon)^2$ . Since these values represent probabilities, the range is constrained to  $\epsilon \in [0,1]$ .

By plotting these output functions (Figure 9), it becomes evident that for one output, the erasure probability decreases, while for the other, it increases.

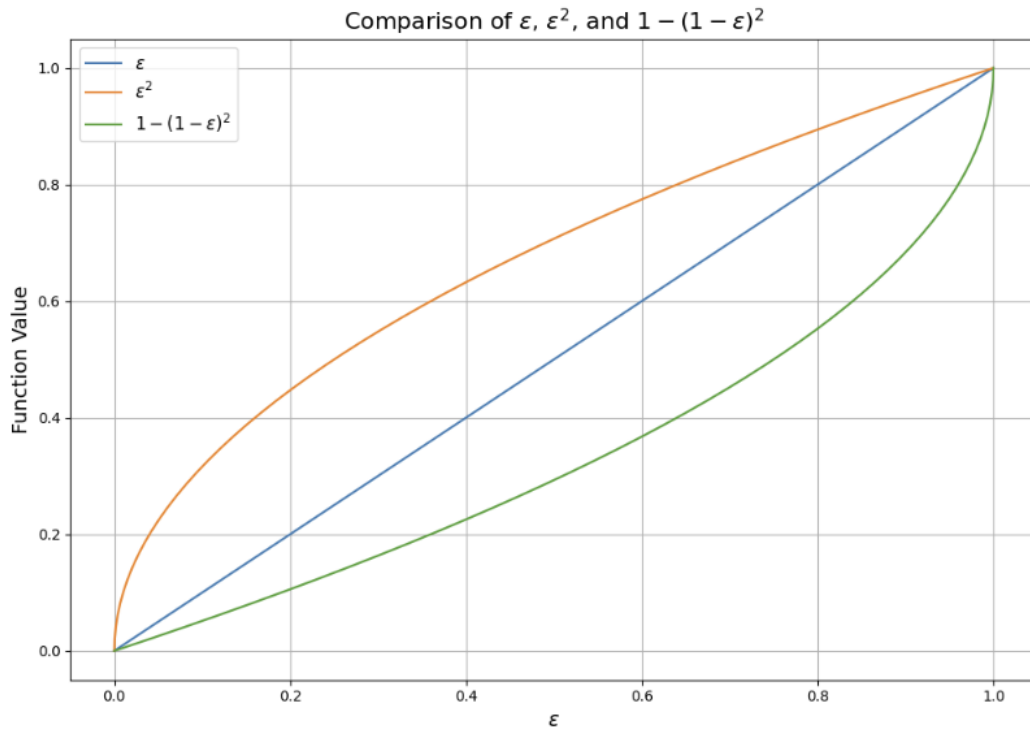


Figure 9: BCD outputs erasure probabilities comparison

In each layer, after passing through the [Bit Channel Decoder Basic Units](#), the new output probabilities are further processed by the [Permutation Block Basic Units](#). These blocks rearrange the probabilities such that the inputs to the next layer's [Bit Channel Decoder Basic Units](#) are grouped into pairs of identical values. This arrangement ensures that, in each layer, every [Bit Channel Decoder Basic Unit](#) produces one output probability higher than its input and one lower. Over successive layers, the probabilities become more polarized, clustering near 0 and 1. Around the midpoint  $\frac{n}{2}$  there is a distinct transition zone. In this region, probabilities gradually shift from being predominantly below 0.5 to predominantly above 0.5, marking the boundary between the two clusters.

### Question 3: Simulate erasure probabilities for different $\epsilon$ values

#### Simulation Results:

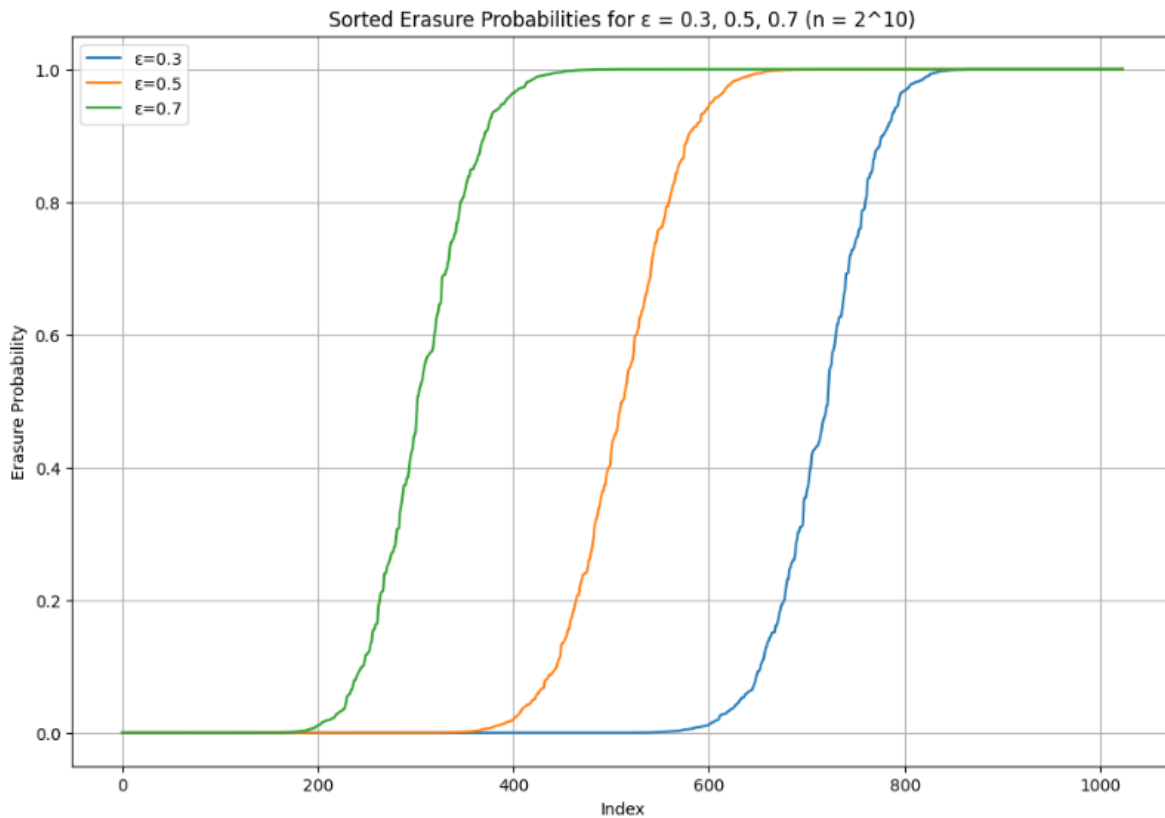


Figure 10: Erasure probabilities simulation for different  $\epsilon$  values and for  $n = 1024$

#### Simulation Analysis:

- As observed in [Question 2](#), we again see dominant clusters for the erasure probability values, concentrated near 0 and 1, with a transition zone whose center depends on the chosen initial value of  $\epsilon$ .
- For a given epsilon value (the initial erasure probability for the input), the boundary point between the clusters (the transition zone) shifts to approximately  $n \cdot (1 - \epsilon)$ .
- Based on the explanation from [Question 2](#) regarding the [U-Transform](#), it makes sense that the initial erasure probability (either 0.3 or 0.7) influences the output. A starting point of 0.3, for example, results in a greater concentration of output samples with lower erasure probabilities.



## Part 2: Encoding, Channeling, and Decoding Analysis

### Question 1: Compute the erasure probabilities for all bit channels

- Let's take the Channel-Decoder part from the given [Encoder-Channel-Decoder Model](#) and annotate it by marking specific points:

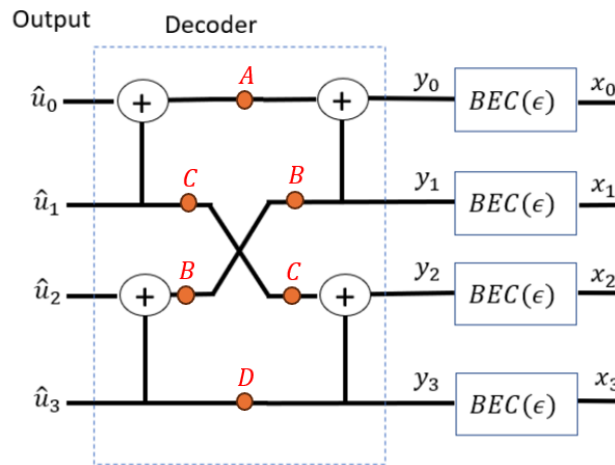


Figure 11: Channel and encoder block from ECD model with additional marks

- From the  $BEC(\epsilon)$  channel definition:  
 $P_e(y_0) = P_e(y_1) = P_e(y_2) = P_e(y_3) = \epsilon$
- From the [Bit Channel Decoder Basic Unit](#) equations:  
 $\epsilon_{AC} = P_e(A | x) = P_e(C | x) = 1 - (1 - \epsilon)^2$   
 $\epsilon_{BD} = P_e(B | x, A) = P_e(D | x, C) = \epsilon^2$
- Applying again the same equation for the decoder output:  
 $P_e(u_0 | x) = 1 - (1 - \epsilon_{AC})^2 = 1 - (1 - \epsilon)^4$   
 $P_e(u_1 | x, u_0) = \epsilon_{AC}^2 = (1 - (1 - \epsilon)^2)^2$   
 $P_e(u_2 | x, u_0, u_1) = 1 - (1 - \epsilon_{BD})^2 = 1 - (1 - \epsilon^2)^2$   
 $P_e(u_3 | x, u_0, u_1, u_2) = \epsilon_{BD}^2 = \epsilon^4$

### Question 2: Writing an encoder function

#### Function Signature:

```
def encode(i: list, info_locs: list, n: int = 4) -> np.ndarray:
```

#### Functions Used:

- transform\_bits\_array:**  
Transforms an input array to an output array using encoder and permutation rules

#### Algorithm:

- Verify if the length of the input information bits vector  $i$  matches the length of the list “info\_locs”.
- Initializing a vector  $u$  with zeros.
- Place the information bits in vector  $u$ .
- Transform  $u$  to the codeword  $x$  using “transform\_bits\_array”

## Question 3: Writing a channel function

### Function Signature:

```
def channel(x: np.ndarray, eps: float) -> np.ndarray:
```

### Algorithm:

- Generate an array called “random\_values” with same length as  $x$  and populate it with random samples from a uniform distribution over  $[0,1)$ .
- Generate a noisy vector  $y$  based on the following rule:

$$y_i = \begin{cases} -1, & \text{random\_value}_i < \epsilon \\ x_i, & \text{else} \end{cases}$$

### Note:

The condition in the generation of the noisy vector is based on the fact that by taking a sample  $s$  from a uniform distribution over  $[0,1)$ , the probability of the event  $\{s < \epsilon\}$  is  $P(\{s < \epsilon\}) = \int_{-\infty}^{\epsilon} f_{s \sim u[0,1]} ds = \int_0^{\epsilon} ds = \epsilon$ .

## Question 4: Writing a decoder function

### Function Signature:

```
def decode(y: np.ndarray, info_locs: list) -> bool:
```

### Functions Used:

- **transform\_by\_equations:**  
Applies the set of equations from the “[Encode-Channel-Decoder Model](#)” on an input array, only for indexes exist in “info\_locs”

### Algorithm:

- Initialize a decoder estimated output vector  $u$  with zeros
- Apply “transform\_by\_equations”, while passing the noisy vector  $y$ , the initialized output vector  $u$ , and the indexes locations “info\_locs”
- Check if any of the elements in the estimated vector  $u$  are erased (represented by  $-1$ )
- Returns the following output:

$$\text{Output} = \begin{cases} \text{True}, & \text{if no erased bits are found} \\ \text{False}, & \text{else} \end{cases}$$

## Question 5: Concatenating encoder-channel-decoder

### Concatenation Implementation:

The concatenation is implemented through a function called “encode\_channel\_decode”, that has the following function signature:

```
def encode_channel_decode(i:list, info_locs:list, eps:float, n:int=4) -> bool:
```

### Simulation:

The simulation for this question can be executed from the attached python file under the title “Question 5: Concatenating encoder-channel-decoder”

## Question 6: Comparison between computed and simulated erasure probability

### Note:

For this question and the subsequent ones, the following functions are utilized:

- **run\_monte\_carlo\_simulation:**  
Perform a Monte Carlo simulation to calculate either the simulated erasure probability or FER for given channel combinations and epsilon values.
- **plot\_simulation\_results:**  
Plot the results of the Monte Carlo simulation and the computed probabilities

### Simulation Results:

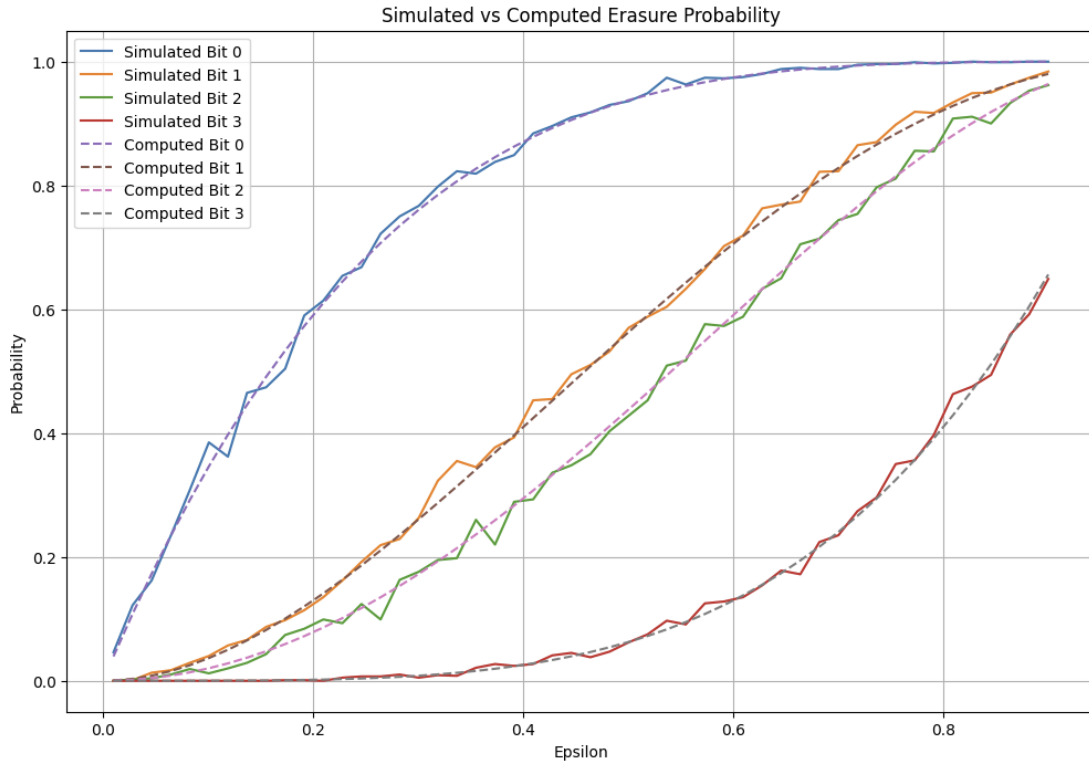


Figure 12: Comparison between computed and simulated erasure probability

- The computed erasure probabilities were calculated using the “computed\_erasure\_probability” function, which is using the equations from [Question 1](#).
- The simulated erasure probabilities were calculated using the “run\_monte\_carlo\_simulation” function, which performs the following steps:
  - For each channel and epsilon value
    - For each Monte Carlo simulation:
      - Generates a random information bit for the current channel
      - Run the “encode\_channel\_decode” function
      - Updating the number of False results from the decoder
    - Calculating the simulated erasure probability by  $P_e = \frac{\#False\ Results}{\#MC\ Simulations}$
- As can be seen from the Plot, the simulated erasure probabilities are aligned with the computed erasure probabilities.

## Question 7: Increasing FER for transmitting in bit channel zero

- Let's define the following sets:
  - $\Omega \equiv \{\omega_j = \{u_j = E\} | j = 0,1,2,3\}$
  - $S \equiv \{\cap_{j \in J} \omega_j | J \subseteq \{0,1,2,3\}, 0 \in J\}$
- Given a system that uses only bit-channel 0 to transmit information, we get  $FER_0 = P(\omega_0)$
- From the dependent of the rest of the u's in  $u_0$ , we know that  $s \subseteq \omega_0$  for each  $s \in S$ , and because also  $\omega_0$  belong to  $S$ , we get
 
$$\bigcup_{s \in S} s = \omega_0$$
- Using the definition of FER, and the fact that bit-channel 0 is always in use, we get
 
$$FER_0 = P(\omega_0) = P\left(\bigcup_{s \in S} s\right) = FER_{transmit-in-all-channels}$$
- To conclude, we can use all channels without increasing the FER.
- The explanation above was confirmed through a simulation of the described case.

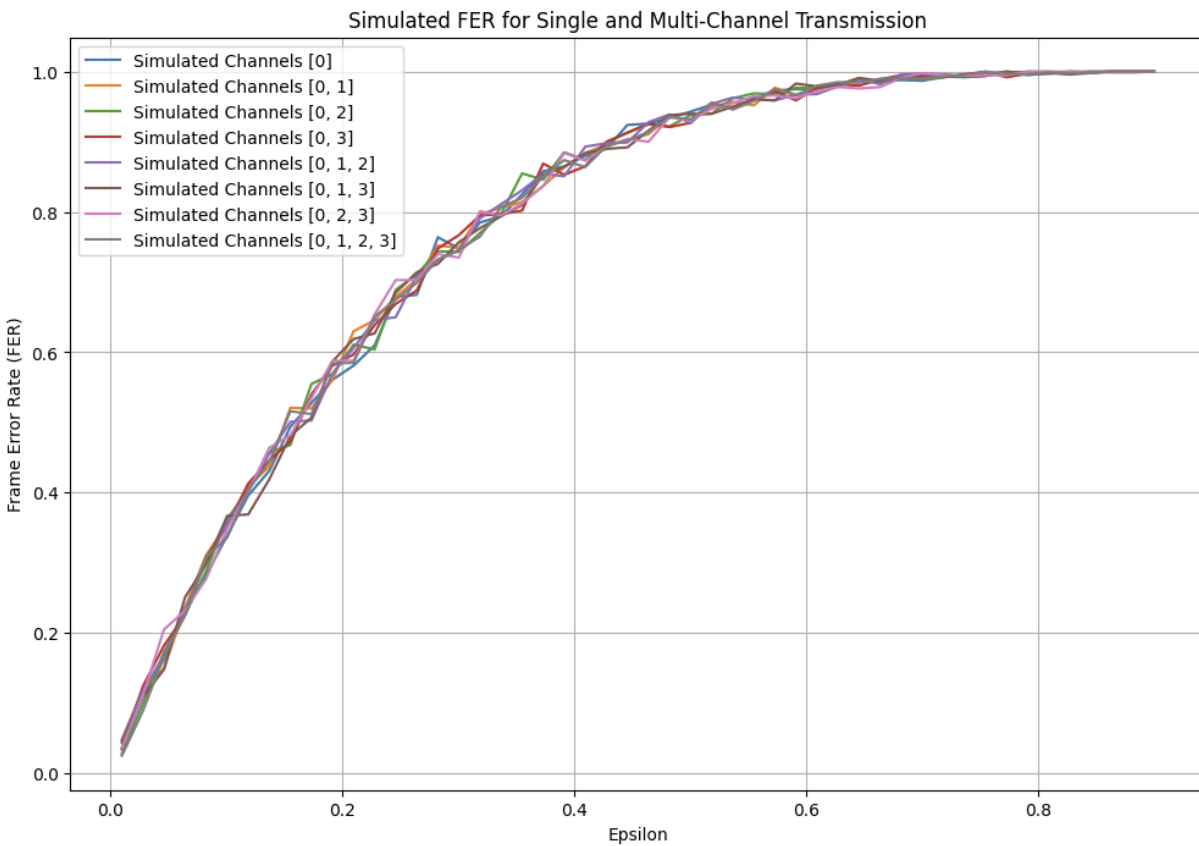


Figure 13: FER simulation for all bit-channels combination that include the zero channel

## Question 8: Selecting the best two bit-channels

Our objective is to identify the two bit-channels with the lowest FER. To achieve this, we perform a FER simulation that evaluates all possible combinations of two bit-channels to determine the optimal pair. The simulation results clearly demonstrate that channels 2 and 3 are the best choice.

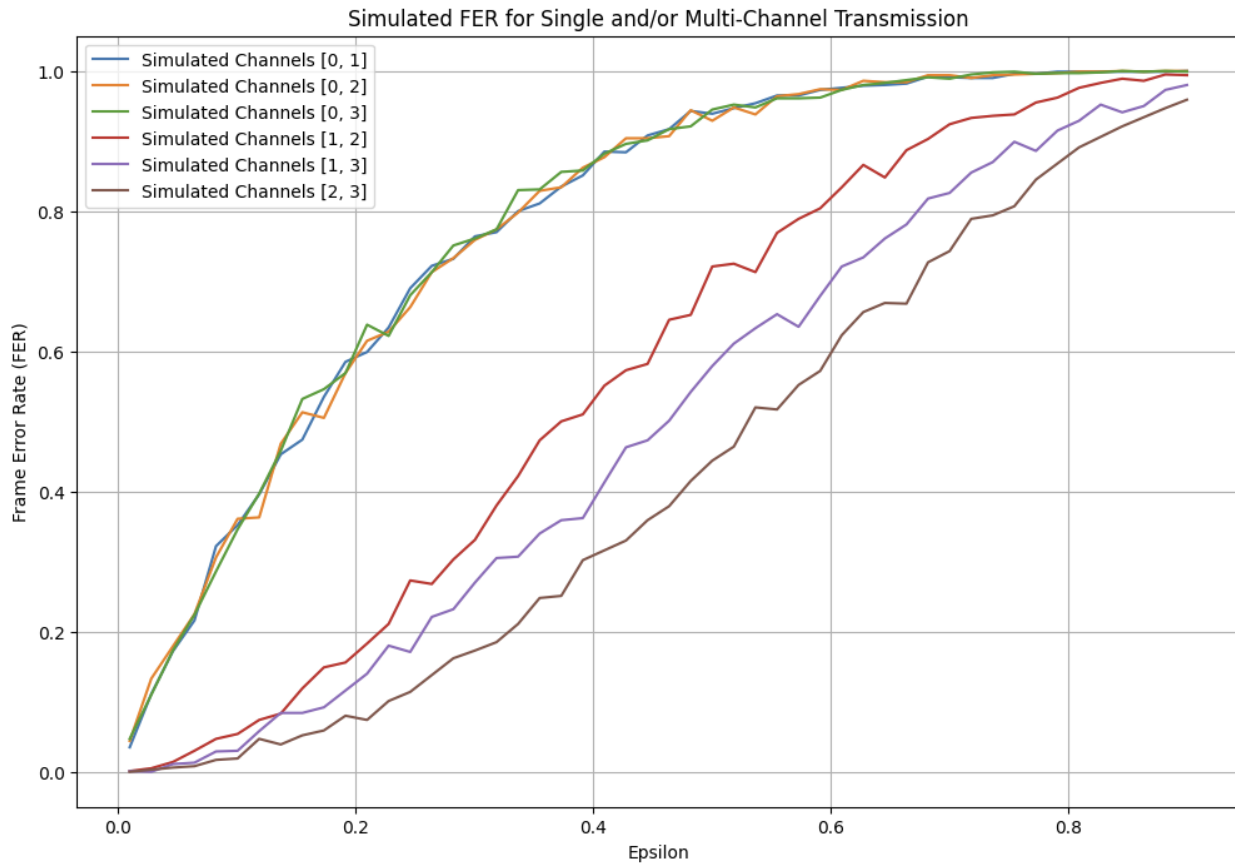


Figure 14: FER simulation for all possible pairs of bit-channel combinations

## Question 9: Computing FER for bit-channels 1 and 2

### Calculating the FER:

$$FER = P(\{u_1 = E\} \cup \{u_2 = E\} | u_0 \text{ given}, x)$$

Using the total probability:

$$(a) FER = P(u_1 = E | u_0 \text{ given}, x) \cdot P(\{u_1 = E\} \cup \{u_2 = E\} | u_0 \text{ given}, u_1 = E, x) + \\ P(u_1 \neq E | u_0 \text{ given}, x) \cdot P(\{u_1 = E\} \cup \{u_2 = E\} | u_0 \text{ given}, u_1 \neq E, x)$$

$$(b) P(\{u_1 = E\} \cup \{u_2 = E\} | u_0 \text{ given}, u_1 = E, x) = P(u_1 = E | u_0, u_1 = E, x) + P(u_2 = E | u_0, u_1 = E, x) - \\ P(u_1 = E | u_0, u_1 = E, x) \cdot P(u_2 = E | u_0, u_1 = E, x) = \\ 1 + P(u_2 = E | u_0, u_1 = E, x) - 1 \cdot P(u_2 = E | u_0, u_1 = E, x) = 1$$

$$(c) P(\{u_1 = E\} \cup \{u_2 = E\} | u_0 \text{ given}, u_1 \neq E, x) = P(u_1 = E | u_0, u_1 \neq E, x) + P(u_2 = E | u_0, u_1 \neq E, x) -$$

$$P(u_1 = E|u_0, u_1 \neq E, x) \cdot P(u_2 = E|u_0, u_1 \neq E, x) =$$

$$0 + P(u_2 = E|u_0, u_1 \neq E, x) - 0 \cdot P(u_2 = E|u_0, u_1 \neq E, x) = P(u_2 = E|u_0, u_1 \neq E, x)$$

From the “[Encoder-Channel-Decoder Model](#)” equation for  $u_2$ :

$$P(u_2 = E|u_0, u_1 \neq E, x) = P_e(y_1 + y_3) \cdot P_e(u_0 + y_0 + y_3) \cdot P_e(u_0 + y_0 + y_2) \cdot P_e(y_1 + y_2)$$

Using again the rule  $P(A \cup B) = P(A) + P(B) - P(A \cap B)$  and the known erasure probabilities from [Question 1](#):

$$P_e(y_1 + y_3) = P_e(y_1 + y_2) = P_e(u_0 + y_0 + y_3) = P_e(u_0 + y_0 + y_2) = 1 - (1 - \epsilon)^2$$

$$\rightarrow (d) \ P(\{u_1 = E\} \cup \{u_2 = E\}|u_0 \text{ given}, u_1 \neq E, x) = P(u_2 = E|u_0, u_1 \neq E, x) = (1 - (1 - \epsilon)^2)^4$$

Substituting (b) and (d) into (a) and using the equation for  $P(u_1 \neq E|u_0 \text{ given}, x)$ :

$$FER = P(u_1 = E|u_0 \text{ given}, x) + P(u_1 \neq E|u_0 \text{ given}, x) \cdot P(\{u_1 = E\} \cup \{u_2 = E\}|u_0 \text{ given}, u_1 \neq E, x) =$$

$$(1 - (1 - \epsilon)^2)^2 + (1 - (1 - (1 - \epsilon)^2)^2) \cdot (1 - (1 - \epsilon)^2)^4$$

### **Simulation Results:**

The simulation results show that the simulated FER aligns closely with the calculated FER curve on the plot.

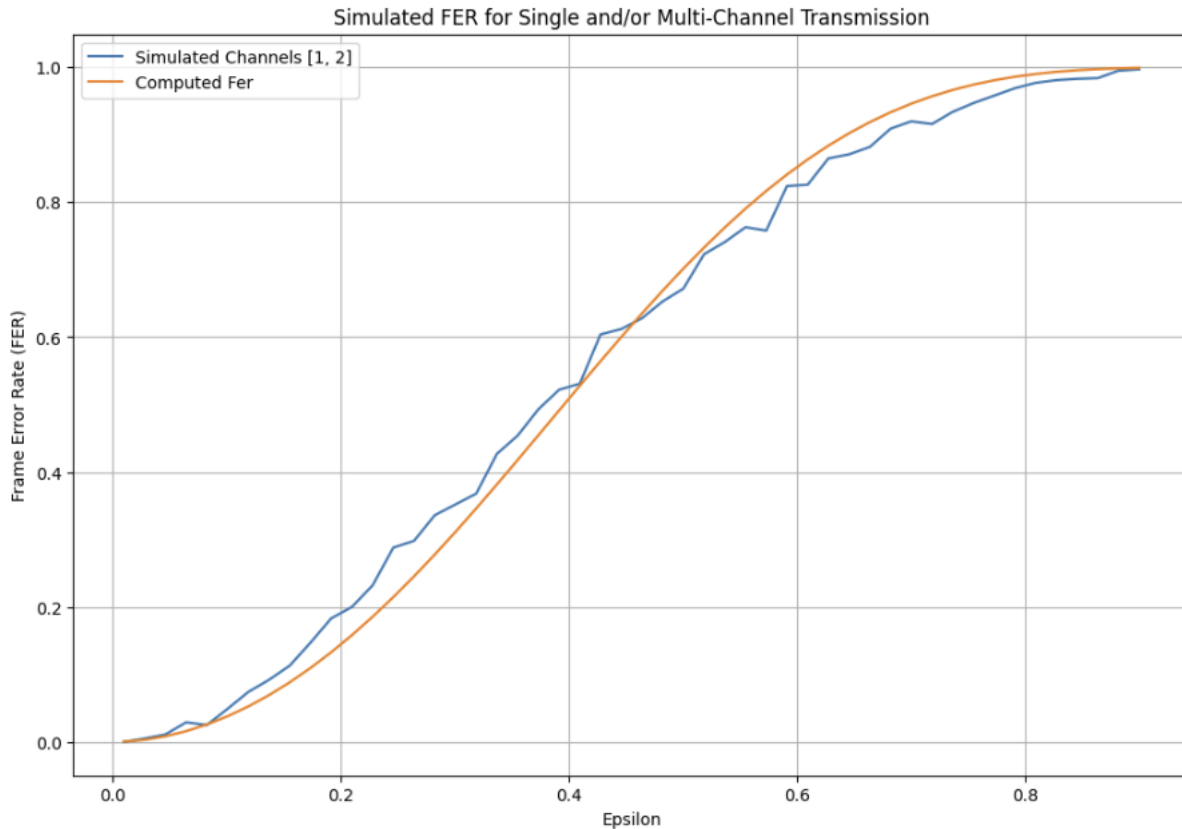


Figure 15: FER simulation for bit-channels 1 and 2