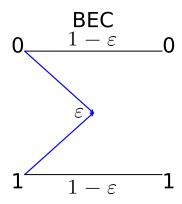
REVIEW

January 20, 2025

Quick Summary of P2P Channels

```
[1]: using Plots, LaTeXStrings
[2]: # Define coordinates for transmitted and received states
     tx = [0, 1]
     rx = [0, 1]
     # Define probabilities as labels
     p = [L"\epsilon", L"1 - \epsilon"];
[3]: # Plot the BSC diagram
     plot(grid=false
         , xaxis=false, yaxis=false
         , framestyle=:none, size = (200,200)
         , title = "BEC"
     plot!([0, 0.5], [1, 0.5], arrow=:arrow, label="", color=:blue) # p line
     plot!([0, 0.5], [0, 0.5], arrow=:arrow, label="", color=:blue) # p line
     plot!([0, 1], [1, 1], label="", color=:black) # 1-p
     plot!([0, 1], [0, 0], label="", color=:black)
     # Annotate the graph
     annotate! (-0.05, 1.05, tx[1]); annotate! (1.05, 1.05, rx[1])
     annotate! (-0.05, -0.05, tx[2]); annotate! (1.05, -0.05, rx[2])
     annotate!(0.4, 0.5, p[1]); annotate!(0.5, -0.1, p[2]); annotate!(0.5, 1.1, p[2])
[3]:
```



1 Binary Erasure Channel (BEC)

- 1. Channel Model:
 - Transmits binary symbols (0 or 1).
 - Each transmitted bit is either:
 - Received correctly with probability 1ϵ , or
 - **Erased** with probability ϵ , represented as an erasure symbol (e).

Example:

- $0 \to 0$ or e,
- $1 \rightarrow 1$ or e.
- 2. Capacity (C): $C = 1 \epsilon$
 - 1: Maximum capacity with no erasures ($\epsilon = 0$).
 - ϵ : Fraction of bits erased by the channel, reducing capacity.
- 3. Behavior:
 - $\epsilon = 0$: Perfect channel, C = 1.
 - $\epsilon = 1$: Completely erasing channel, C = 0.
 - For $0 < \epsilon < 1$: Capacity decreases linearly as ϵ increases.

Compact Intuition: The Binary Erasure Channel (BEC) capacity is the fraction of bits successfully transmitted. Erasures (ϵ) reduce capacity by removing information from the channel.

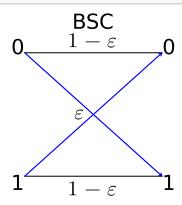
```
[4]: # Plot the BSC diagram
plot(grid=false
    , xaxis=false, yaxis=false
    , framestyle=:none, size = (200,200)
    , title = "BSC"
)

plot!([0, 1], [1, 0], arrow=:arrow, label="", color=:blue) # p line
plot!([0, 1], [0, 1], arrow=:arrow, label="", color=:blue) # p line
plot!([0, 1], [1, 1], label="", color=:black) # 1-p
```

```
plot!([0, 1], [0, 0], label="", color=:black)

# Annotate the graph
annotate!(-0.05, 1.05, tx[1]); annotate!(1.05, 1.05, rx[1])
annotate!(-0.05, -0.05, tx[2]); annotate!(1.05, -0.05, rx[2])
annotate!(0.4, 0.5, p[1]); annotate!(0.5, -0.1, p[2]); annotate!(0.5, 1.1, p[2])
```

[4]:



2 Binary Symmetric Channel (BSC)

1. Channel Model:

- Transmits binary symbols (0 or 1).
- Each bit has a probability ϵ of being flipped.
- Error probability: $P(0 \to 1) = P(1 \to 0) = \epsilon$.
- Correct transmission probability: $P(0 \to 0) = P(1 \to 1) = 1 \epsilon$.

2. Capacity (C_{BSC}) : $C_{BSC} = 1 - H_2(\epsilon)$

- 1: Maximum capacity without errors.
- $H_2(\epsilon)$: Binary entropy function, representing uncertainty due to errors.

3. Binary Entropy Function $(H_2(\epsilon))$: $H_2(\epsilon) = -\epsilon \cdot \log_2(\epsilon) - (1-\epsilon) \cdot \log_2(1-\epsilon)$

- $H_2(0) = 0$: No errors, full capacity.
- $H_2(0.5) = 1$: Maximum uncertainty, no capacity.

4. Behavior:

- $\epsilon = 0$: Perfect channel, $C_{BSC} = 1$.
- $\epsilon = 0.5$: Completely noisy, $C_{BSC} = 0$.
- For $0 < \epsilon < 0.5$: Capacity decreases as ϵ increases.

Compact Intuition: The BSC capacity is the theoretical maximum rate of reliable data transmission, reduced by the uncertainty caused by errors. Lower ϵ means higher capacity, while higher ϵ reduces it.

2.0.1 AWGN Channel Summary

- 1. Channel Model: y = x + w, $w \sim N(0, N_0)$
 - x: Transmitted signal.
 - y: Received signal.
 - w: Gaussian noise with zero mean and variance N_0 the noise power spectral density.
- 2. Signal Power: $P = E[|x|^2]$
- 3. Signal-to-Noise Ratio (SNR): $SNR = \frac{P}{N_0}$
- 4. Channel Capacity: $C = \log_2{(1 + SNR)} = \log_2{\left(1 + \frac{P}{N_0}\right)}$
 - C: Maximum achievable data rate (in bps/Hz).
- 5. **Key Behavior**:
 - $P \uparrow$ (high signal power): $C \uparrow$ (more capacity).
 - $N_0 \uparrow$ (high noise): $C \downarrow$ (less capacity).

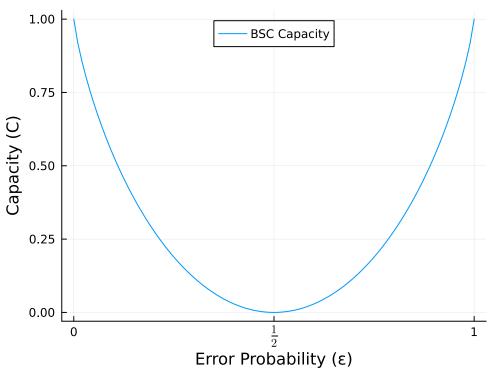
Insight: The AWGN channel capacity quantifies the theoretical limit of reliable communication over a noisy channel.

```
[5]: using Plots
     # Define the binary entropy function
    function H()
         if == 0 || == 1
             return 0.0
        end
        return - * log2() - (1 - ) * log2(1 - )
    end
    # Define the BSC capacity function
    function C ()
        return 1 - H()
    end
     # Generate values of from 0 to 1
     _values = 0:0.01:1
    capacities = C .(_values)
     # Plot the capacity curve
    plot(_values, capacities,
         label = "BSC Capacity"
         , xlabel = "Error Probability ()", ylabel = "Capacity (C)"
         , title = "Binary Symmetric Channel (BSC) Capacity"
```

```
legend = :top, size = (500,400)
xticks = (0:0.5:1, ["0", L"\frac{1}{2}", "1"])
```

[5]:

Binary Symmetric Channel (BSC) Capacity



Communication System Components:

\$

(2)

$$\boxed{\text{Sink}} \leftarrow \hat{x} = (\hat{x}_1, \dots, \hat{x}_n) \leftarrow \boxed{\text{Decoder}} \leftarrow y = (y_1, \dots, y_n) \qquad \boxed{}$$

\$

- 1. Source: Produces the information $\underline{x} = (x_1, \dots, x_n)$.
- 2. **Encoder**: Transforms \underline{x} into a codeword $\underline{c} = (c_1, \dots, c_n)$, adding redundancy.
- 3. Channel: Transmits \underline{c} , introducing errors, resulting in $\underline{y}=(y_1,\dots,y_n).$
- 4. **Decoder**: Processes \underline{y} to estimate $\hat{\underline{x}} = (\hat{x}_1, \dots, \hat{x}_n)$, correcting errors.
- 5. **Sink**: Receives $\hat{\underline{x}}$, ideally matching \underline{x} .

2.1.1 Objective:

Ensure $\hat{x} = x$ despite channel errors.

2.1.2 Properties of Linear Block Code

- 1. Linearity:
 - The set of codewords \mathcal{X} forms a linear subspace of \mathbb{F}_2^n .
 - Any linear combination of codewords is also a valid codeword: $\underline{v}_1 + \underline{v}_2 \in \mathcal{X}, \ \forall \underline{v}_1, \underline{v}_2 \in \mathcal{X}.$
- 2. Generator Matrix (G):
 - The $k \times n$ generator matrix G maps k-bit input vectors $(\underline{u} \in \mathbb{F}_2^k)$ to n-bit codewords $(v = u^{\top}G).$
 - Defines the structure of the codebook \mathcal{X} .
- 3. Code Rate (*R*):
 - The ratio of information bits to total bits: $R = \frac{k}{n}$
 - Indicates the efficiency of the code.
- 4. Code Size ($|\mathcal{X}|$):
 - The number of unique codewords: $|\mathcal{X}| = 2^k$
- 5. Minimum Hamming Distance (d_{\min}) :
 - The smallest Hamming distance between any two distinct codewords.
 - Determines the error-detecting and error-correcting capability: $t=\left|\frac{d_{\min}-1}{2}\right|$
 - t: Maximum correctable errors.
 - $-d_{\min}-1$: Maximum detectable errors.
- 6. Parity Check Matrix (H):
 - The H matrix defines the null space of G: $HG^{\top} = 0$
 - Used to verify codewords: $H\underline{v}^{\top} = 0 \implies \underline{v} \in \mathcal{X}$.
- 7. Error Detection and Correction:
 - Error detection: Capable of detecting up to $d_{\min} 1$ errors. Error correction: Can correct up to $\lfloor \frac{d_{\min} 1}{2} \rfloor$ errors.
- 8. Redundancy:
 - The number of redundant bits added for error correction is n-k, where n is the codeword length.

2.1.3 Compact Summary:

- Linear subspace: Codewords form a subspace of \mathbb{F}_2^n .
- Size: $|\mathcal{X}| = 2^k$.
- Rate: $R = \frac{k}{n}$.
- Error capabilities: Based on d_{\min} .
- Defined by:
 - Generator matrix (G).
 - Parity check matrix (H).

Error Correction Error Correction Code \mathcal{X}

Linear error corr. Code \mathcal{X}

Code Rate:

Explanation:

• Code Rate (R) is the fraction of a codeword used for information:

- k: Number of information bits.
- -n: Total number of bits in the codeword (information + redundancy).

Trade-off:

- Higher R ($R \to 1$): More efficient but less error protection.
- Lower R $(R \to 0)$: Less efficient but better error correction.

 $\mathcal{X} = \{\underline{v} = \underline{u}^{\top}G, \ \underline{u} \in \mathbb{F}_2^k\}$ Linear Block Code Definition:

Explanation:

• \mathcal{X} : The set of all codewords in the code (codebook).

• \underline{u} : A binary **message vector** of length k ($\underline{u} \in \mathbb{F}_2^k$).

• G: The generator matrix $(k \times n)$ used to map u to a codeword.

• \underline{v} : A binary **codeword** of length n ($\underline{v} \in \mathbb{F}_2^n$).

Key Points:

1. Each \underline{u} maps to a unique \underline{v} via $\underline{v} = \underline{u}^{\top}G$.

2. The total number of codewords is 2^k (one for each u).

3. \mathcal{X} is a **linear subspace** of dimension k in \mathbb{F}_2^n .

Compact Summary: \mathcal{X} is the codebook of a linear block code, where each codeword \underline{v} is generated by multiplying a k-bit message vector u with the generator matrix G.

• Generator Matrix: G

Code Size: $|\mathcal{X}| = 2^k$

- k-bit message vectors $(\underline{u} \in \mathbb{F}_2^k)$ are mapped to n-bit codewords $(\underline{v} = \underline{u}^\top G)$ via the generator matrix G.
- The codebook \mathcal{X} forms a linear subspace with 2^k unique codewords, corresponding to the k-bit input combinations.

7

• Minimum Hamming Distance:

 $d_{\min} = \min\{d_H(x_i, x_i)\}, \ \forall x_i, x_i \in \mathcal{X}, \ x_i \neq x_i \ (\text{Hamming distance})$

• Linear Code Minimum Distance: $d_{\min} = \min\big(\tfrac{x_i, \underline{x_j} \in \mathcal{X}}{x_i \neq x_j}\big)\big\{d_H(x_i, x_j)\big\} = \min x \in \mathcal{X}\big\{W_H(\underline{x})\big\}$

• Parity Check Relation: $\underline{v}H^\top = 0 \implies \underline{v} \in \mathcal{X} \subseteq \mathbb{F}_2^n$

2.1.4 Parity Check Matrix

```
• H \in \mathbb{F}_2^{(n-k) \times n}
```

- $\dim(\operatorname{Im}(G)) = k$
- $H = \text{null}(G^{\top}), \text{ dim} = n k$

```
[7]: # Message vector

u = [1 0 1 0]

# Generate codeword

v = mod.(u * G, 2)

println("Generated codeword: ", v) # Should be a valid codeword
```

Generated codeword: [1 0 1 0 1 0 1]

```
[8]: # Parity check validation
parity_check = mod.(v * H', 2)
println("Parity check result: ", parity_check) # Should be [0, 0, 0]
```

Parity check result: [0 0 0]

```
[9]: using LinearAlgebra

# Define the generator matrix G (4x7 for (7,4) code)

G = [
    1 0 0 0 1 1 0;
    0 1 0 0 1 0 1;
    0 0 1 0 0 1 1;
    0 0 0 1 1 1 1
```

```
# Generate all possible messages (4-bit binary combinations)
messages = [bitstring(i)[end-3:end] for i in 0:2^4-1] # 4-bit binary strings
u = [parse.(Int, split(m, ""))' for m in messages]; @show u; # Convert to row⊔
 \neg vectors (1x4) u \setminus underbar
# Generate all codewords using G
  = [mod.(u * G, 2) for u in \underline{u}]; @show
# Define a function to compute Hamming distance
function hamming_distance(v , v)
    sum(v .!= v) # Count differing elements
end
# Compute the minimum Hamming distance using a comprehension
d = minimum(
    hamming_distance([i], [j]) for (i, j)
         in Iterators.product(1:length(), 1:length()) if i < j</pre>
)
# Output the result
println("Minimum distance of the code: ", d )
<u>u</u> = Adjoint{Int64, Vector{Int64}}[[0 0 0 0], [0 0 0 1], [0 0 1 0], [0 0 1 1],
```

u = Adjoint{Int64, Vector{Int64}}[[0 0 0 0], [0 0 0 1], [0 0 1 0], [0 0 1 1],
[0 1 0 0], [0 1 0 1], [0 1 1 0], [0 1 1 1], [1 0 0 0], [1 0 0 1], [1 0 1 0], [1
0 1 1], [1 1 0 0], [1 1 0 1], [1 1 1 0], [1 1 1 1]]
= [[0 0 0 0 0 0 0], [0 0 0 1 1 1 1], [0 0 1 0 0 1 1], [0 0 1 1 1 0 0], [0 1 0
0 1 0 1], [0 1 0 1 0 1 0], [0 1 1 0 1 1 0], [0 1 1 1 0 0 1], [1 0 0 0 1 1], [1 0 1 0 1 0], [1
0 0 1 0 0 1], [1 0 1 0 1 0 1], [1 0 1 1 0 1 0], [1 1 0 0 0 1 1], [1 1 0 1 1 0
0], [1 1 1 0 0 0 0], [1 1 1 1 1 1]]
Minimum distance of the code: 3