

# Lecture 4: Capacity

## Project info

1. Each project group consists of two students.
2. Each project group should as soon as possible, send an email to [fredrik.rusek@eit.lth.se](mailto:fredrik.rusek@eit.lth.se) and [mg7107ma-s@student.lu.se](mailto:mg7107ma-s@student.lu.se) containing Name and email address to each project member.'
3. The project group should contact Fredrik Rusek to decide about project and articles!
4. Each group should write a project report.
5. The structure of the project report should follow journal articles published by IEEE. However, two columns are not needed.
6. The project report should be written in English *with your own words, tables and figures*, and contain 4-5 pages.
7. The report should be clearly written, and written to the other students in this course!
8. Observe copyright rules: "copy & paste" is in general strictly forbidden!

# Lecture 4: Capacity

## Project info

9. Book a meeting with Mgeni at your earliest possible convenience in case you are interested in discussing how a good report should be written (book meetings with him via email). You can also discuss topics and articles with Mgeni
10. The project report should be sent in .pdf format to Mgeni before Wednesday 12 December, 17.00
11. Feedback on the reports will be provided in a meeting with Mgeni (book meetings with him via email)
12. Oral presentations in the week starting with Monday December 17
13. Each group should have relevant comments and questions on the project report and on the oral presentation of another group. NOTE! The project presentation should be clear and aimed to the other students in this course! After the oral presentation the project report and the presentation will be discussed (5 min).
14. Final report should be sent to Fredrik and Mgeni at latest January 11, 2019.

# Lecture 4: Capacity

## Power efficiency

We know from before (e.g., union bound) that  $P_s \leq cQ \left( \sqrt{d_{\min}^2 \frac{E_b}{N_0}} \right)$

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Now, divide both sides with the bandwidth  $W$   $\frac{R_b}{W} \leq \frac{\mathcal{P}}{N_0 W} \frac{d_{\min}^2}{\mathcal{X}}$



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We have seen this before, it is defined as...

Now, divide both sides with the bandwidth  $W$   $\left( \frac{R_b}{W} \right) \leq \frac{\mathcal{P}}{N_0 W} \frac{d_{\min}^2}{\mathcal{X}}$

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We have seen this before, it is defined as bandwidth efficiency

Now, divide both sides with the bandwidth  $W$   $\rho = \frac{R_b}{W} \leq \frac{\mathcal{P}}{N_0 W} \frac{d_{\min}^2}{\mathcal{X}}$

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Now, divide both sides with the bandwidth  $W$   $\rho \leq \frac{\mathcal{P}}{N_0 W} \frac{d_{\min}^2}{\mathcal{X}}$  Power efficiency

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Now, divide both sides with the bandwidth  $W$   $\rho \leq \frac{\mathcal{P}}{N_0 W} \frac{d_{\min}^2}{\mathcal{X}}$  Performance req

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Bandwidth and power efficiencies are linked

Now, divide both sides with the bandwidth  $W$   $\rho \leq \frac{\mathcal{P}}{N_0 W} \frac{d_{\min}^2}{\mathcal{X}}$

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Unit ?

Now, divide both sides with the bandwidth  $W$

$$\rho \leq \frac{\mathcal{P}}{N_0 W} \frac{d_{\min}^2}{\mathcal{X}}$$

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Power

Now, divide both sides with the bandwidth  $W$

$$\rho \leq \frac{\mathcal{P}}{N_0 W} \frac{d_{\min}^2}{\mathcal{X}} \quad \frac{W}{W}$$

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Bandwidth

Now, divide both sides with the bandwidth  $W$   $\rho \leq \frac{\mathcal{P}}{N_0 W} \frac{d_{\min}^2}{\mathcal{X}}$   $\frac{W}{\text{Hz}}$



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Spectral density

Now, divide both sides with the bandwidth  $W$

$$\rho \leq \frac{\mathcal{P}}{N_0 W} \frac{d_{\min}^2}{\mathcal{X}}$$

$$\frac{W}{? \text{ Hz}}$$

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Spectral density

Now, divide both sides with the bandwidth  $W$

$$\rho \leq \frac{\mathcal{P}}{N_0 W} \frac{d_{\min}^2}{\mathcal{X}}$$

$$\frac{W}{W/\text{Hz Hz}}$$

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Has no unit (dimensionless)

Now, divide both sides with the bandwidth  $W$

$$\rho \leq \frac{\mathcal{P}}{N_0 W} \frac{d_{\min}^2}{\mathcal{X}}$$

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Dito

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Received signal-to-noise-power-ratio

Now, divide both sides with the bandwidth  $W$

$$\rho \leq \frac{\mathcal{P}}{N_0 W} \frac{d_{\min}^2}{\mathcal{X}}$$

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Definition

Now, divide both sides with the bandwidth  $W$

$$\rho \leq \mathcal{SNR}_r \frac{d_{\min}^2}{\mathcal{X}}$$

# Lecture 4: Capacity

## Power efficiency

"BW efficiency" = "Signal-to-noise-power-ratio"  $\times$  "Power efficiency"

$$\rho \leq \mathcal{SNR}_r \frac{d_{\min}^2}{\mathcal{X}}$$



# Lecture 4: Capacity

## Shannon Capacity

Before going on, we go through what the term capacity means

Given a scalar channel of form  $y = \sqrt{A}x + n$ ,  $n \sim CN(0, N_0)$

We know that the capacity is  $C = \log_2 \left( 1 + \frac{A}{N_0} \right)$

But what does this mean?


# Lecture 4: Capacity

## Shannon Capacity

$$y = \sqrt{A}x + n, \quad n \sim CN(0, N_0)$$
$$C = \log_2 \left( 1 + \frac{A}{N_0} \right)$$

Build a codebook of all information sequences possible to send of length  $K$

000000	.....	00
000000	.....	01
000000	.....	10
111111	.....	10
111111	.....	11



$K$

# Lecture 4: Capacity

## Shannon Capacity

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Build a codebook of all information sequences possible to send of length K

000000	.....	00
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111111	.....	10
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K

Sending K bits of information means:

pick one of the rows, and tell the receiver which row it is

# Lecture 4: Capacity

## Shannon Capacity

$$y = \sqrt{A}x + n, \quad n \sim CN(0, N_0)$$
$$C = \log_2 \left( 1 + \frac{A}{N_0} \right)$$

Information book

000000 ..... 00  
000000 ..... 01  
000000 ..... 10

111111 ..... 10  
111111 ..... 11

K

Build a codebook of codewords to send for each information word, length N

$x_{11}x_{12}x_{13}x_{14} \dots x_{1(N-1)}x_{1N}$

$x_{21}x_{22}x_{23}x_{24} \dots x_{2(N-1)}x_{2N}$

$x_{2^K 1}x_{2^K 2}x_{2^K 3}x_{2^K 4} \dots x_{2^K(N-1)}x_{2^K N}$

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## Shannon Capacity

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### Information book

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000000 ..... 01  
000000 ..... 10

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111111 ..... 11

K

### Codebook

$x_{11}x_{12}x_{13}x_{14} \dots x_{1(N-1)}x_{1N}$   
 $x_{21}x_{22}x_{23}x_{24} \dots x_{2(N-1)}x_{2N}$

$x_{2^K 1}x_{2^K 2}x_{2^K 3}x_{2^K 4} \dots x_{2^K(N-1)}x_{2^K N}$

N

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## Shannon Capacity

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$$C = \log_2 \left( 1 + \frac{A}{N_0} \right)$$

### Information book

000000	.....	00
000000	.....	01
000000	.....	10

If this is my data

111111	.....	10
111111	.....	11

K

### Codebook

$x_{11}x_{12}x_{13}x_{14}$	.....	$x_{1(N-1)}x_{1N}$
$x_{21}x_{22}x_{23}x_{24}$	.....	$x_{2(N-1)}x_{2N}$

$x_{2^k1}x_{2^k2}x_{2^k3}x_{2^k4}$	.....	$x_{2^k(N-1)}x_{2^kN}$
------------------------------------	-------	------------------------

N

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### Information book

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If this is my data

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111111	.....	11

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### Codebook

$x_{11}x_{12}x_{13}x_{14}$	.....	$x_{1(N-1)}x_{1N}$
$x_{21}x_{22}x_{23}x_{24}$	.....	$x_{2(N-1)}x_{2N}$

I send this one

$x_{2^k1}x_{2^k2}x_{2^k3}x_{2^k4}$	.....	$x_{2^k(N-1)}x_{2^kN}$
------------------------------------	-------	------------------------

N

# Lecture 4: Capacity

## Shannon Capacity

As  $x$  over this channel used  $N$  times

$$y = \sqrt{A}x + n, \quad n \sim CN(0, N_0)$$

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$K$

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$x_{2^k 1}x_{2^k 2}x_{2^k 3}x_{2^k 4}$	.....	$x_{2^k(N-1)}x_{2^k N}$
--	-------	-------------------------

$N$



# Lecture 4: Capacity

## Shannon Capacity

$$y = \sqrt{A}x + n, \quad n \sim CN(0, N_0)$$
$$C = \log_2 \left( 1 + \frac{A}{N_0} \right)$$

Clearly, bit rate is  $K/N$  bits/channel use

### Information book

000000 ..... 00  
000000 ..... 01  
000000 ..... 10

111111 ..... 10  
111111 ..... 11

K

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$x_{2^K 1}x_{2^K 2}x_{2^K 3}x_{2^K 4} \dots x_{2^K(N-1)}x_{2^K N}$

N

# Lecture 4: Capacity

Receiver observes

$y_1 y_2 y_3 y_4 \dots y_{(N-1)} y_N$

$$y = \sqrt{A}x + n, \quad n \sim CN(0, N_0)$$
$$C = \log_2 \left( 1 + \frac{A}{N_0} \right)$$

Information book

000000 ..... 00  
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Codebook

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# Lecture 4: Capacity

Receiver observes

$y_1 y_2 y_3 y_4 \dots y_{(N-1)} y_N$

Compare with this one

$$d_1 = \sum_{n=1}^N |y_n - x_{1n}|^2$$

Information book

000000 ..... 00  
000000 ..... 01  
000000 ..... 10

111111 ..... 10  
111111 ..... 11

K

Codebook

$x_{11} x_{12} x_{13} x_{14} \dots x_{1(N-1)} x_{1N}$   
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$$y = \sqrt{A}x + n, \quad n \sim CN(0, N_0)$$

$$C = \log_2 \left( 1 + \frac{A}{N_0} \right)$$

# Lecture 4: Capacity

Receiver observes

$y_1 y_2 y_3 y_4 \dots y_{(N-1)} y_N$

Compare with this one

$$d_2 = \sum_{n=1}^N |y_n - x_{2n}|^2$$

Information book

000000 ..... 00  
000000 ..... 01  
000000 ..... 10

111111 ..... 10  
111111 ..... 11

K

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# Lecture 4: Capacity

Receiver observes

$y_1 y_2 y_3 y_4 \dots y_{(N-1)} y_N$

Compare with this one

$$d_{2^K} = \sum_{n=1}^N |y_n - x_{2^K n}|^2$$

Information book

000000 ..... 00  
000000 ..... 01  
000000 ..... 10

111111 ..... 10  
111111 ..... 11

K

Codebook

$x_{11} x_{12} x_{13} x_{14} \dots x_{1(N-1)} x_{1N}$   
 $x_{21} x_{22} x_{23} x_{24} \dots x_{2(N-1)} x_{2N}$

$x_{2^K 1} x_{2^K 2} x_{2^K 3} x_{2^K 4} \dots x_{2^K (N-1)} x_{2^K N}$

N

$$y = \sqrt{A}x + n, \quad n \sim CN(0, N_0)$$

$$C = \log_2 \left( 1 + \frac{A}{N_0} \right)$$

# Lecture 4: Capacity

Receiver observes

$y_1 y_2 y_3 y_4 \dots y_{(N-1)} y_N$

Take smallest

$$d_2 = \sum_{n=1}^N |y_n - x_{2n}|^2$$

Information book

000000 ..... 00  
000000 ..... 01  
000000 ..... 10

111111 ..... 10  
111111 ..... 11

K

$$y = \sqrt{A}x + n, \quad n \sim CN(0, N_0)$$

$$C = \log_2 \left( 1 + \frac{A}{N_0} \right)$$

Codebook

$x_{11} x_{12} x_{13} x_{14} \dots x_{1(N-1)} x_{1N}$   
 $x_{21} x_{22} x_{23} x_{24} \dots x_{2(N-1)} x_{2N}$

$x_{2^k 1} x_{2^k 2} x_{2^k 3} x_{2^k 4} \dots x_{2^k(N-1)} x_{2^k N}$

N

# Lecture 4: Capacity

Receiver observes

$y_1 y_2 y_3 y_4 \dots y_{(N-1)} y_N$

Take smallest

$$d_2 = \sum_{n=1}^N |y_n - x_{2n}|^2$$

Information book

000000	.....	00
000000	.....	01
000000	.....	10

So data is this one

111111	.....	10
111111	.....	11

K

$$y = \sqrt{A}x + n, \quad n \sim CN(0, N_0)$$

$$C = \log_2 \left( 1 + \frac{A}{N_0} \right)$$

Codebook

$x_{11} x_{12} x_{13} x_{14} \dots x_{1(N-1)} x_{1N}$
$x_{21} x_{22} x_{23} x_{24} \dots x_{2(N-1)} x_{2N}$

$x_{2^k 1} x_{2^k 2} x_{2^k 3} x_{2^k 4} \dots x_{2^k(N-1)} x_{2^k N}$
--

N

# Lecture 4: Capacity

Receiver observes

$y_1 y_2 y_3 y_4 \dots y_{(N-1)} y_N$

Take smallest

$$d_2 = \sum_{n=1}^N |y_n - x_{2n}|^2$$

Information book

000000	.....	00
000000	.....	01
000000	.....	10

So data is this one

111111	.....	10
111111	.....	11

K

This is ML decoding and is optimal

Capacity means the following

Codebook

$x_{11} x_{12} x_{13} x_{14} \dots x_{1(N-1)} x_{1N}$
$x_{21} x_{22} x_{23} x_{24} \dots x_{2(N-1)} x_{2N}$

$x_{2^k 1} x_{2^k 2} x_{2^k 3} x_{2^k 4} \dots x_{2^k(N-1)} x_{2^k N}$
--

N



# Lecture 4: Capacity

Receiver observes

$Y_1 Y_2 Y_3 Y_4 \dots Y_{(N-1)} Y_N$

Take smallest

$$d_2 = \sum_{n=1}^N |y_n - x_{2n}|^2$$

Information book

000000	.....	00
000000	.....	01
000000	.....	10

So data is this one

111111	.....	10
111111	.....	11

K

This is ML decoding and is optimal

Capacity means the following

1. If  $K/N \leq C$ , and  $K \rightarrow \infty$  then  $\text{Prob}(\text{Correct detection}) = 1$

Codebook

$x_{11} x_{12} x_{13} x_{14} \dots x_{1(N-1)} x_{1N}$
$x_{21} x_{22} x_{23} x_{24} \dots x_{2(N-1)} x_{2N}$

$x_{2^k 1} x_{2^k 2} x_{2^k 3} x_{2^k 4} \dots x_{2^k(N-1)} x_{2^k N}$
--

N

# Lecture 4: Capacity

Receiver observes

$Y_1 Y_2 Y_3 Y_4 \dots Y_{(N-1)} Y_N$

Take smallest

$$d_2 = \sum_{n=1}^N |y_n - x_{2n}|^2$$

Information book

000000	.....	00
000000	.....	01
000000	.....	10

So data is this one

111111	.....	10
111111	.....	11

K

This is ML decoding and is optimal

Capacity means the following

1. If  $K/N \leq C$ , and  $K \rightarrow \infty$  then  
Prob(Correct detection)=1
2. If  $K/N > C$ , then  
Prob(Incorrect detection)=1

Codebook

$x_{11} x_{12} x_{13} x_{14} \dots x_{1(N-1)} x_{1N}$
$x_{21} x_{22} x_{23} x_{24} \dots x_{2(N-1)} x_{2N}$

$x_{2^k 1} x_{2^k 2} x_{2^k 3} x_{2^k 4} \dots x_{2^k(N-1)} x_{2^k N}$
--

N

# Lecture 4: Capacity

Receiver observes

$y_1 y_2 y_3 y_4 \dots y_{(N-1)} y_N$

Take smallest

$$d_2 = \sum_{n=1}^N |y_n - x_{2n}|^2$$

Information book

000000	.....	00
000000	.....	01
000000	.....	10

So data is this one

111111	.....	10
111111	.....	11

K

To reach  $C$ , code-symbols must be  
Random complex Gaussian variables  
That is, generate codebook randomly

If it is generated with, say, 16QAM  
 $C$  cannot be reached

Codebook

$x_{11} x_{12} x_{13} x_{14} \dots x_{1(N-1)} x_{1N}$
$x_{21} x_{22} x_{23} x_{24} \dots x_{2(N-1)} x_{2N}$

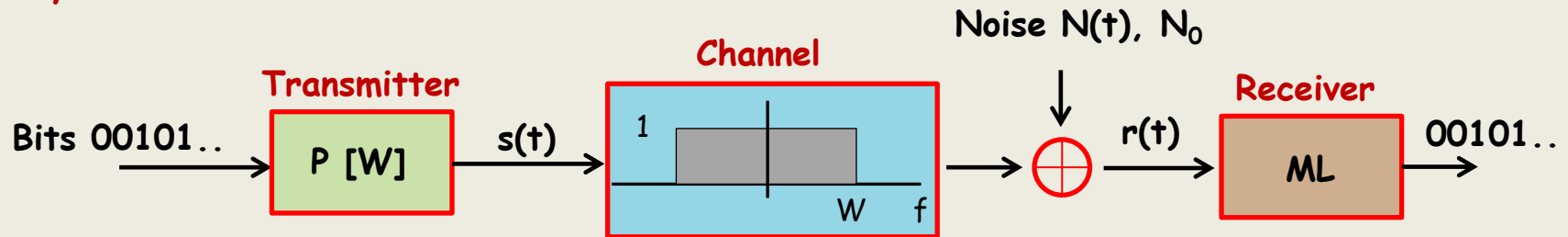
$x_{2^k 1} x_{2^k 2} x_{2^k 3} x_{2^k 4} \dots x_{2^k(N-1)} x_{2^k N}$
--

N

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

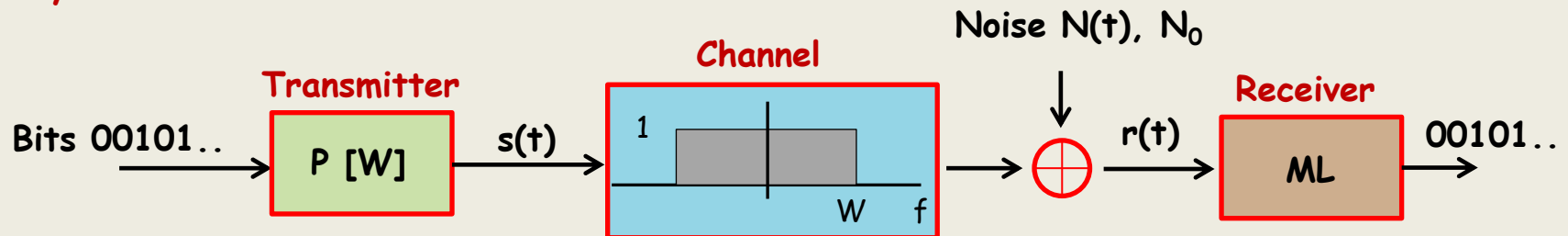
System model:



# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



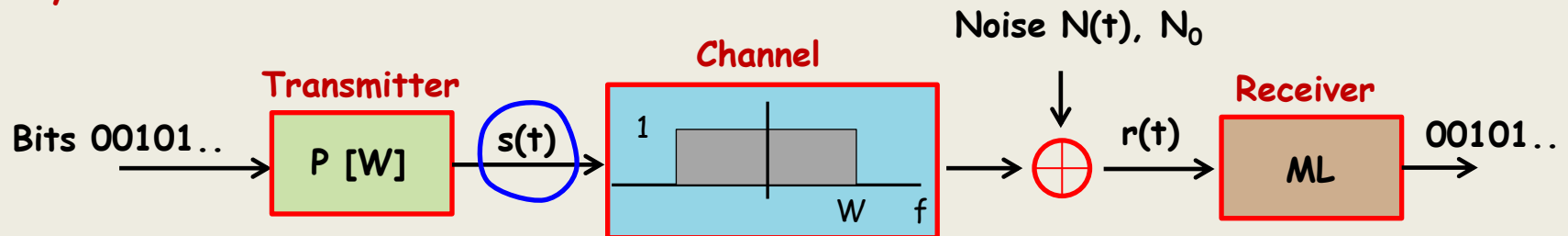
Interpretation of capacity:

Given a transmission of length  $T$  (seconds)

# Lecture 4: Capacity

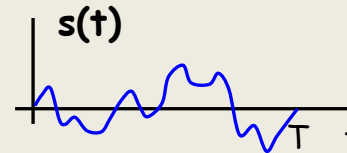
## Extension to continuous channel (Shannon '48)

System model:



Interpretation of capacity:

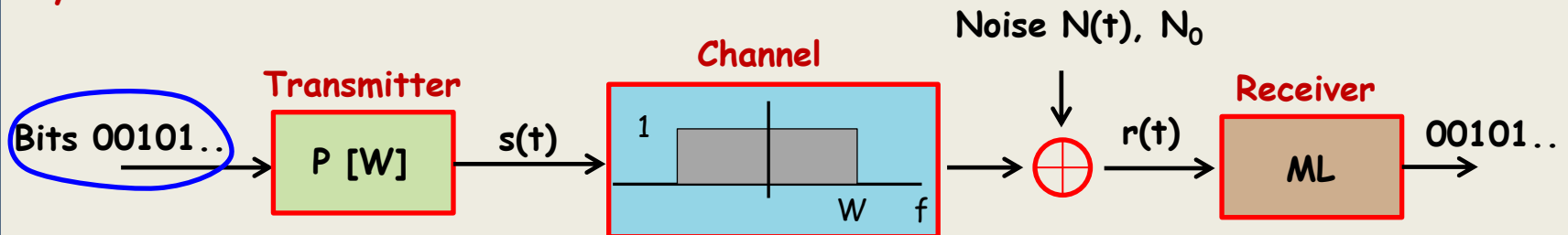
Given a transmission of length  $T$  (seconds)



# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

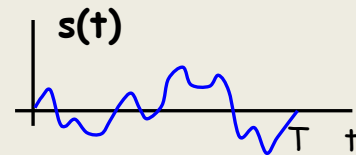
System model:



Interpretation of capacity:

Given a transmission of length  $T$  (seconds)

And a number of bits  $K$



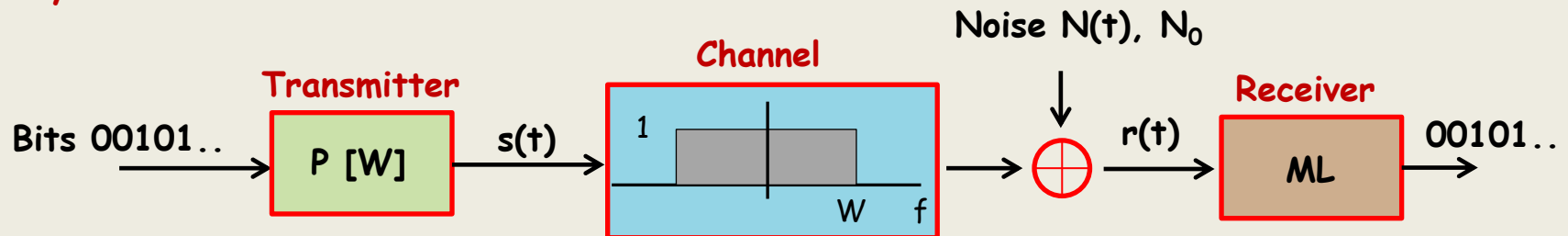
Bits 0010111010110100...010011

A blue bracket underneath the bit sequence is labeled  $K$ , indicating the total number of bits transmitted.

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:

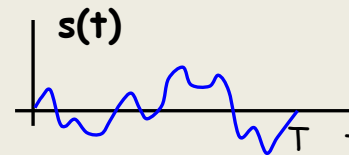


Interpretation of capacity:

Given a transmission of length  $T$  (seconds)

And a number of bits  $K$

The bitrate is:  $K/T$  [bit/sec]



Bits 0010111010110100...010011

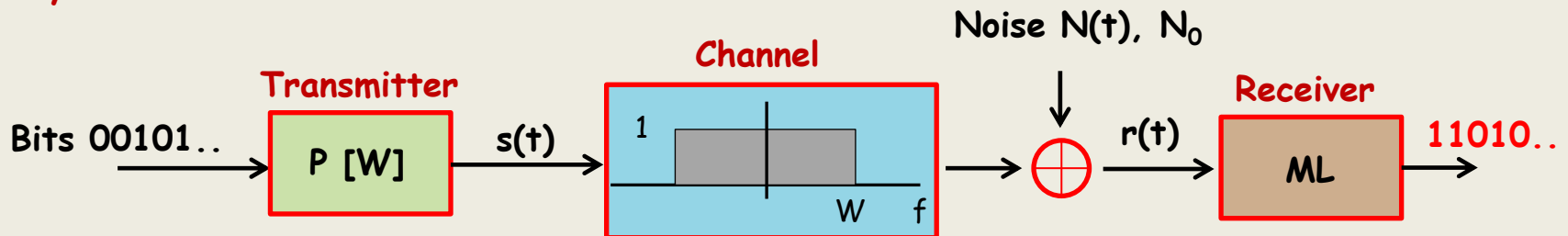
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# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



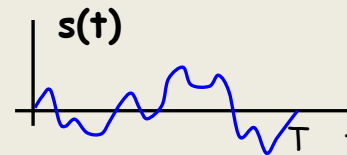
Interpretation of capacity:

Given a transmission of length  $T$  (seconds)

And a number of bits  $K$

The bitrate is:  $K/T$  [bit/sec]

If  $K/T$  is too high, then many errors



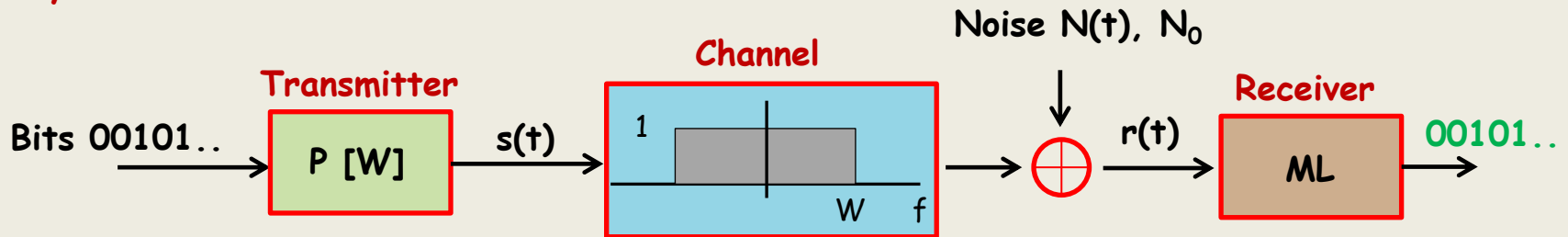
Bits 0010111010110100...010011

A blue bracket underneath the bit sequence is labeled  $K$ , indicating the total number of bits.

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



Interpretation of capacity:

Given a transmission of length  $T$  (seconds)

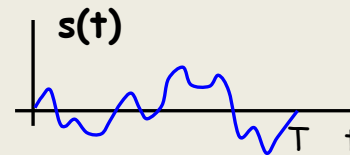
And a number of bits  $K$

The bitrate is:  $K/T$  [bit/sec]

If  $K/T$  is too high, then many errors

Shannon proved: Possible to have NO ERRORS if,

1)  $T \rightarrow \infty$     2)  $\lim_{T \rightarrow \infty} \frac{K}{T} = C = W \log_2 \left( 1 + \frac{P}{N_0 W} \right)$



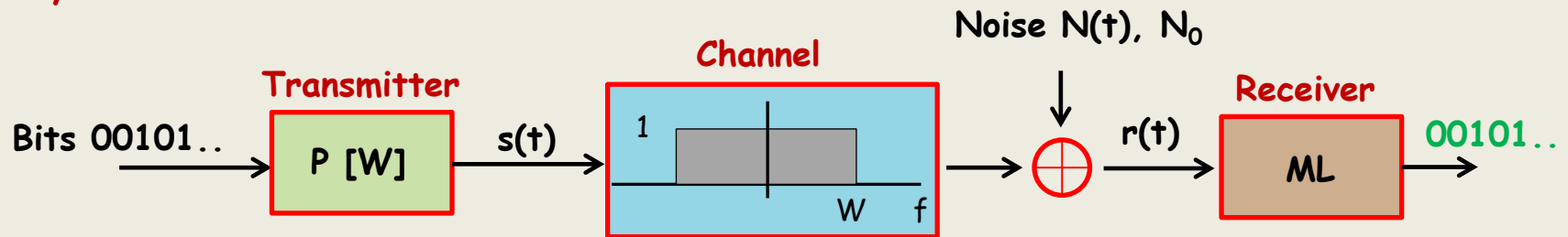
Bits 0010111010110100...010011

A blue bracket underneath the first 16 bits of the sequence is labeled  $K$ , indicating the total number of bits transmitted.

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



$$C = W \log_2 \left( 1 + \frac{P}{N_0 W} \right)$$

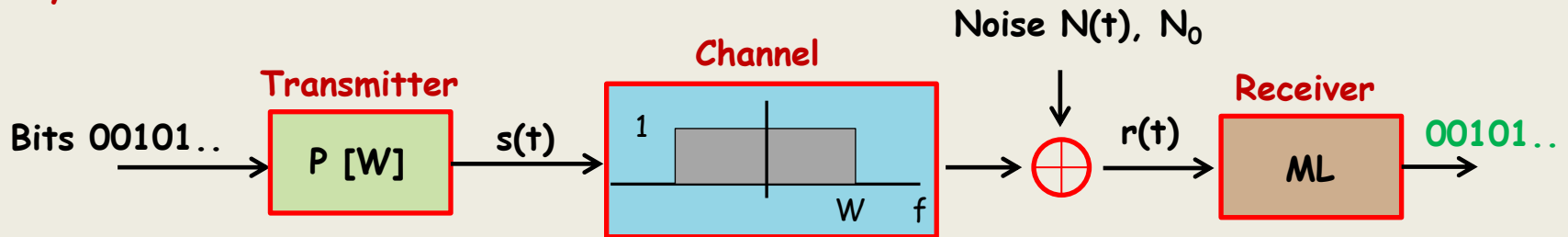
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1.  $C$  is not power, nor bandwidth efficiency  
( $C$  is not dimensionless)

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



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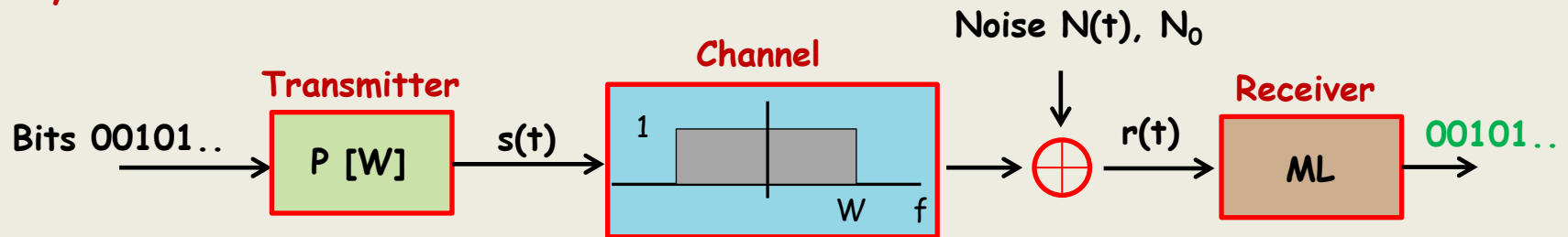
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2. Not easy to reach  $C$   
(i.e., to find a set of  $s(t)$  signals)

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



$$C = W \log_2 \left( 1 + \frac{P}{N_0 W} \right)$$

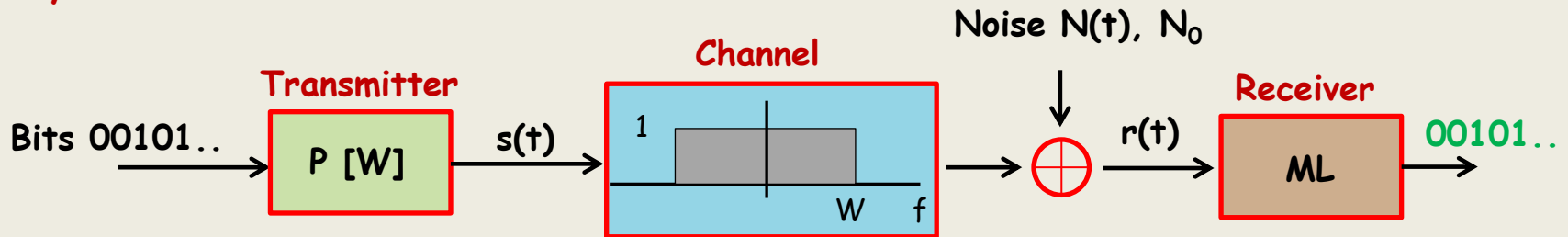
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3. There is no parameter called  $d_{\min}^2$

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



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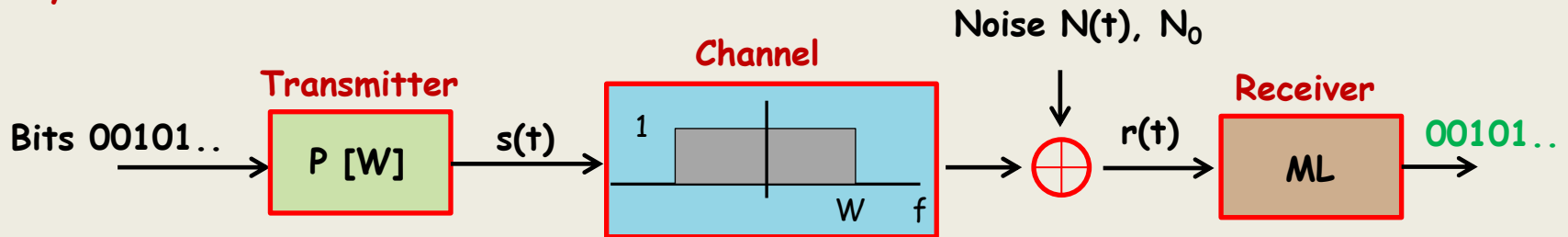
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3. There is no parameter called  $d_{\min}^2$
4. When  $W$  grows:

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



$$C = W \log_2 \left( 1 + \frac{P}{N_0 W} \right)$$

Grows linearly

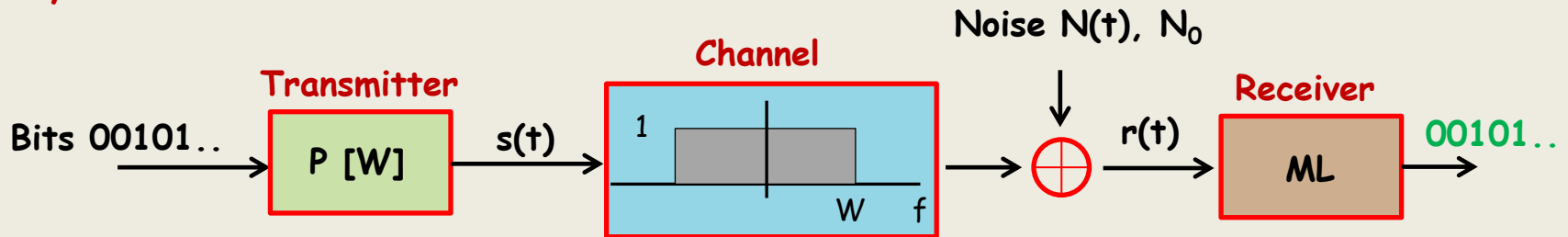
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3. There is no parameter called  $d_{\min}^2$
4. When  $W$  grows:

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



$$C = W \log_2 \left( 1 + \frac{P}{N_0 W} \right)$$

Grows linearly      decreases

**Facts:**

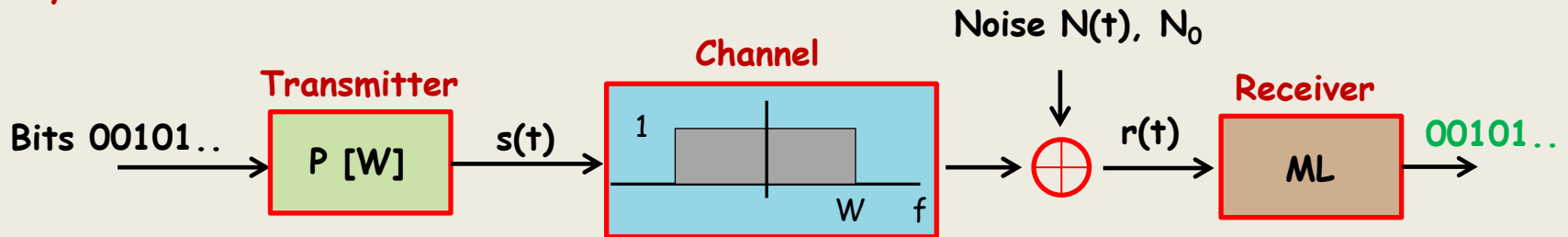
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2. Not easy to reach  $C$   
(i.e., to find a set of  $s(t)$  signals)
3. There is no parameter called  $d_{\min}^2$
4. When  $W$  grows:



# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



Decreases logarithmically

$$C = W \log_2 \left( 1 + \frac{P}{N_0 W} \right)$$

Grows linearly

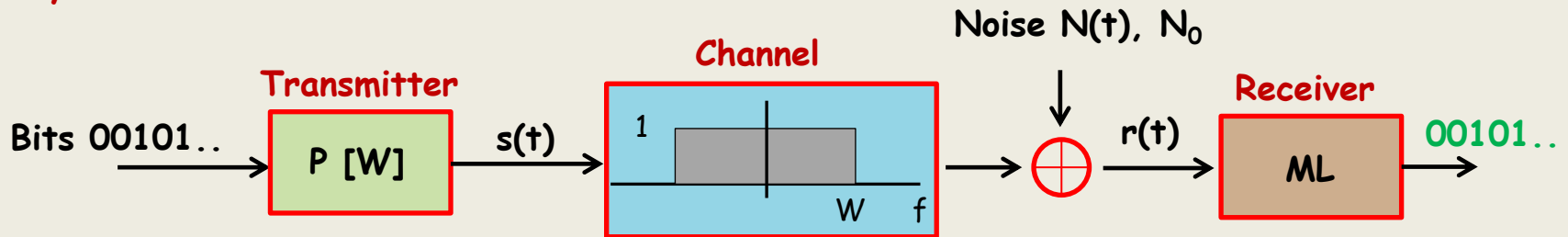
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3. There is no parameter called  $d_{\min}^2$
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# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



$$C = W \log_2 \left( 1 + \frac{P}{N_0 W} \right)$$

Grows

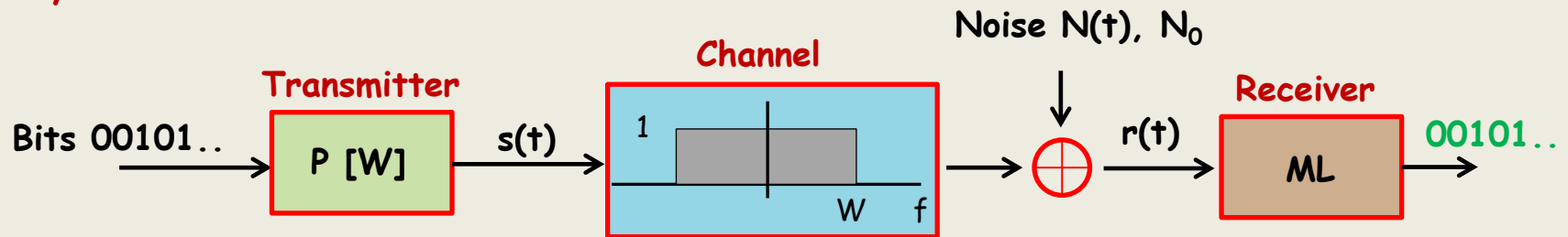
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2. Not easy to reach  $C$   
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3. There is no parameter called  $d_{\min}^2$
4. When  $W$  grows:  $C$  grows

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

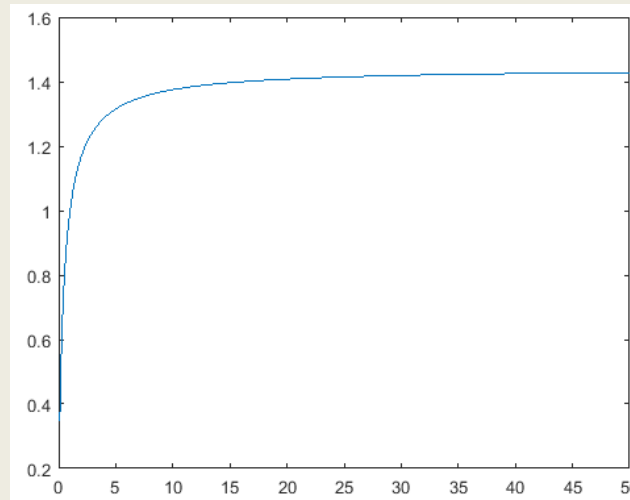
System model:



$$C = W \log_2 \left( 1 + \frac{P}{N_0 W} \right)$$

Example,  $P/N_0 = 1$

$C$



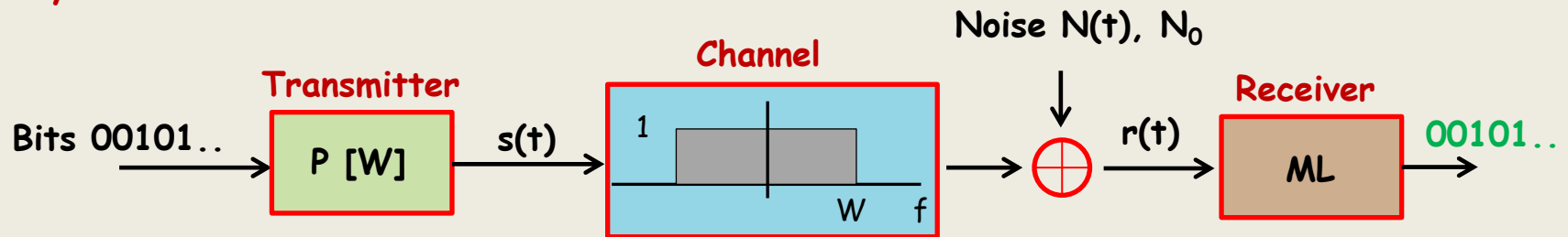
$W$

But it grows to a limit

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



$$C = W \log_2 \left( 1 + \frac{P}{N_0 W} \right)$$

What is the limit?

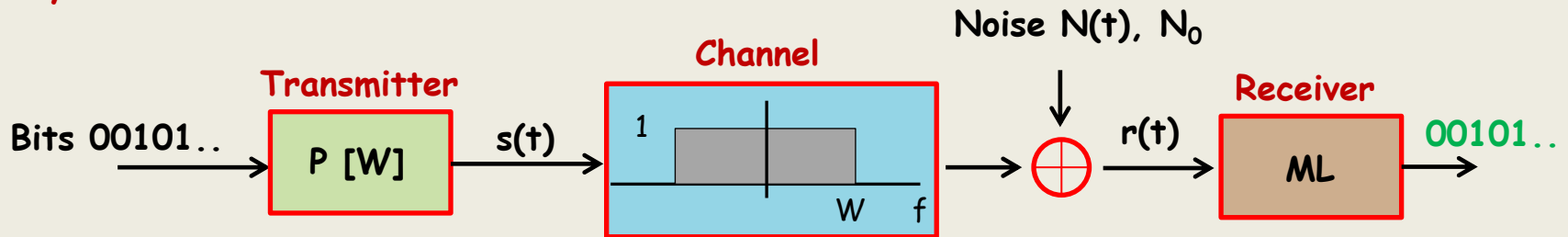
Standard limit

$$\lim_{x \rightarrow \infty} x \ln \left( 1 + \frac{A}{x} \right) = A$$

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



$$C = W \log_2 \left( 1 + \frac{P}{N_0 W} \right)$$

What is the limit?

Standard limit

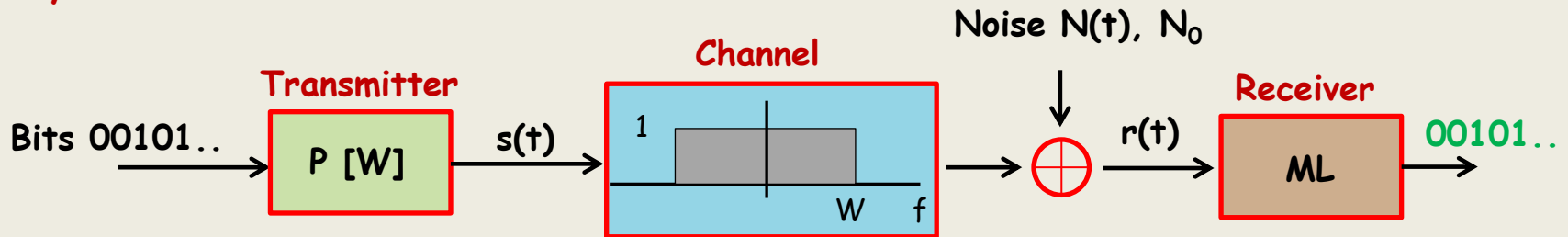
Identify  $x$  with  $W$

$$\lim_{x \rightarrow \infty} x \ln \left( 1 + \frac{A}{x} \right) = A$$

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



$$C = W \log_2 \left( 1 + \frac{P}{N_0 W} \right)$$

What is the limit?

Standard limit

Identify  $x$  with  $W$

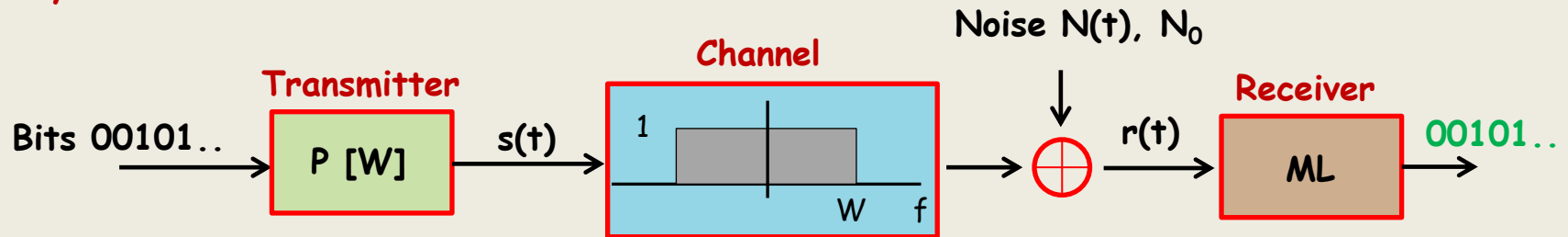
Identify  $A$  with  $P/N_0$

$$\lim_{x \rightarrow \infty} x \ln \left( 1 + \frac{A}{x} \right) = A$$

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



$$C = \frac{W}{\ln(2)} \ln \left( 1 + \frac{P}{N_0 W} \right)$$

What is the limit?

Standard limit

Identify  $x$  with  $W$

Identify  $A$  with  $P/N_0$

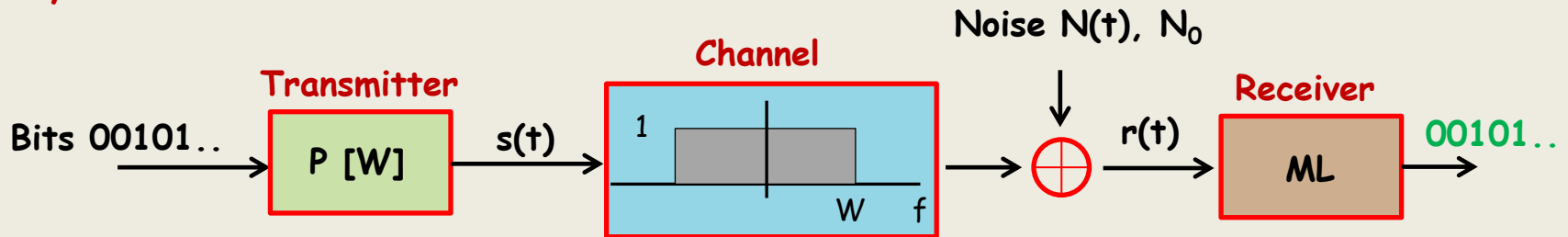
Express  $\log_2(x)$  as  $\ln(x)/\ln(2)$

$$\lim_{x \rightarrow \infty} x \ln \left( 1 + \frac{A}{x} \right) = A$$

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



$$C = \frac{W}{\ln(2)} \ln \left( 1 + \frac{P}{N_0 W} \right)$$

What is the limit?

Standard limit

$$\lim_{x \rightarrow \infty} x \ln \left( 1 + \frac{A}{x} \right) = A$$

Carry out limit

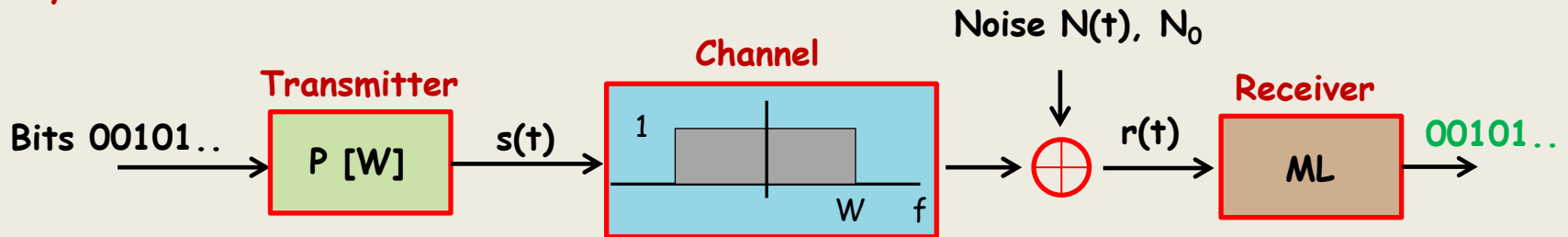
$$C_{\max} = \lim_{W \rightarrow \infty} \frac{W}{\ln(2)} \ln \left( 1 + \frac{P}{N_0 W} \right) =$$



# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



$$C = \frac{W}{\ln(2)} \ln \left( 1 + \frac{\mathcal{P}}{N_0 W} \right)$$

What is the limit?

Standard limit

$$\lim_{x \rightarrow \infty} x \ln \left( 1 + \frac{A}{x} \right) = A$$

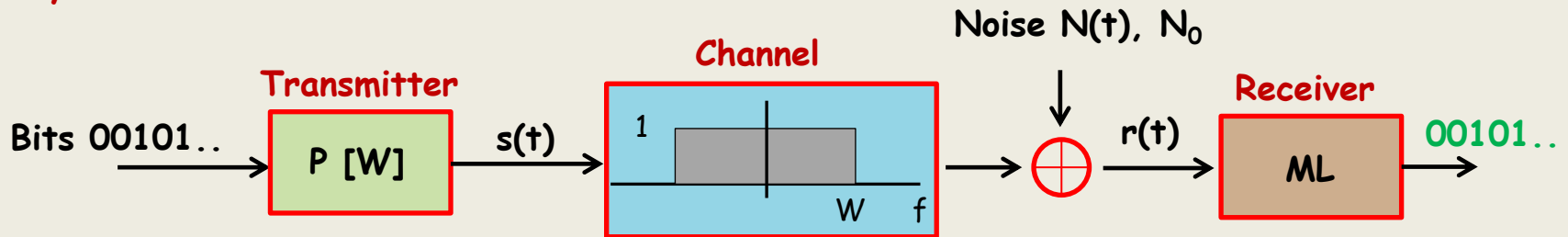
Carry out limit

$$C_{\max} = \lim_{W \rightarrow \infty} \frac{W}{\ln(2)} \ln \left( 1 + \frac{\mathcal{P}}{N_0 W} \right) = \frac{\mathcal{P}}{N_0 \ln(2)}$$

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



$$C = \frac{W}{\ln(2)} \ln \left( 1 + \frac{P}{N_0 W} \right)$$

What is the limit?

Standard limit

$$\lim_{x \rightarrow \infty} x \ln \left( 1 + \frac{A}{x} \right) = A$$

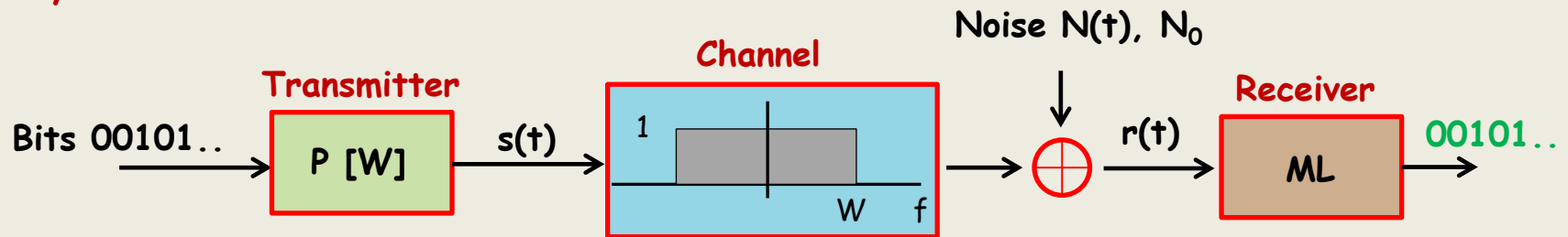
Carry out limit

$$C_{\max} = \lim_{W \rightarrow \infty} \frac{W}{\ln(2)} \ln \left( 1 + \frac{P}{N_0 W} \right) = \frac{P}{N_0 \ln(2)}$$

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:

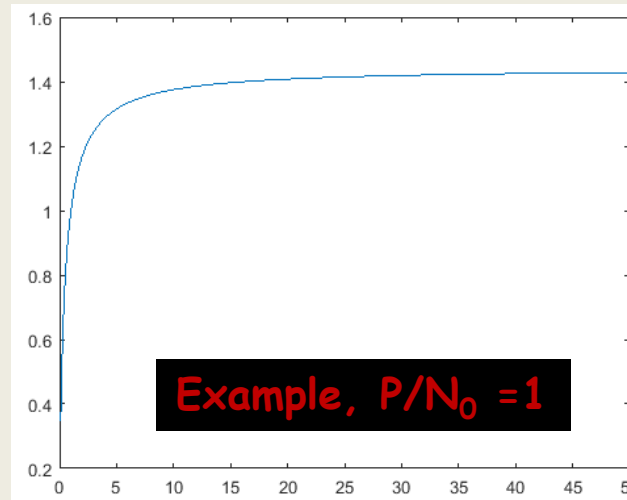


$$C = \frac{W}{\ln(2)} \ln \left( 1 + \frac{P}{N_0 W} \right)$$

Standard limit

$$\lim_{x \rightarrow \infty} x \ln \left( 1 + \frac{A}{x} \right) = A$$

**C**



But it grows to a limit

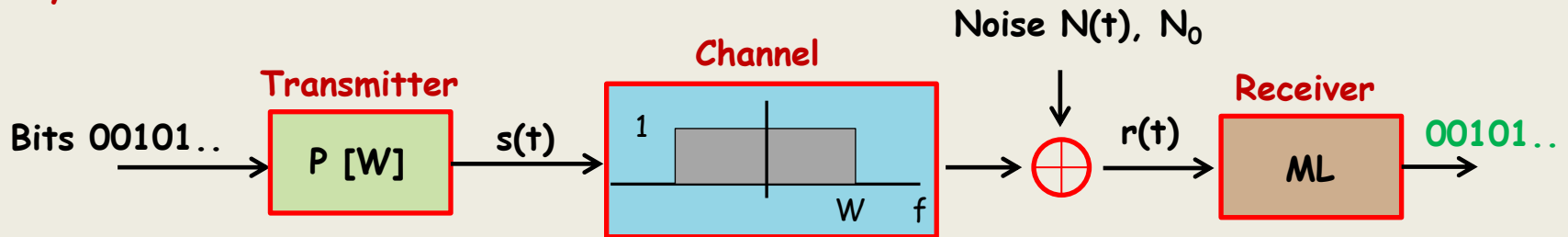
$$C_{\max} = \frac{P}{N_0 \ln(2)}$$

**W**

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:

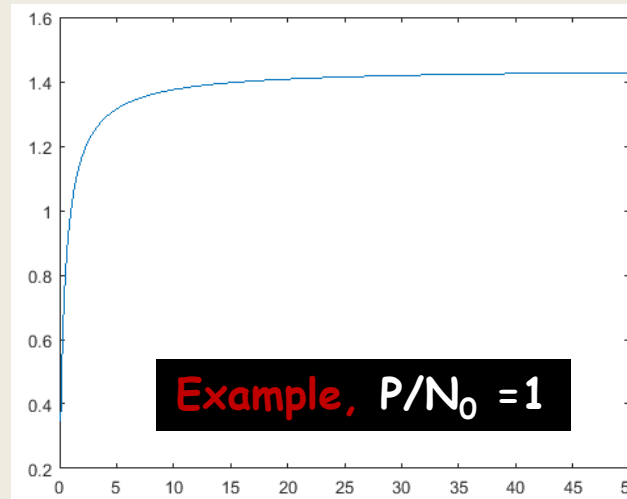


$$C = \frac{W}{\ln(2)} \ln \left( 1 + \frac{P}{N_0 W} \right)$$

Standard limit

$$\lim_{x \rightarrow \infty} x \ln \left( 1 + \frac{A}{x} \right) = A$$

**C**



But it grows to a limit

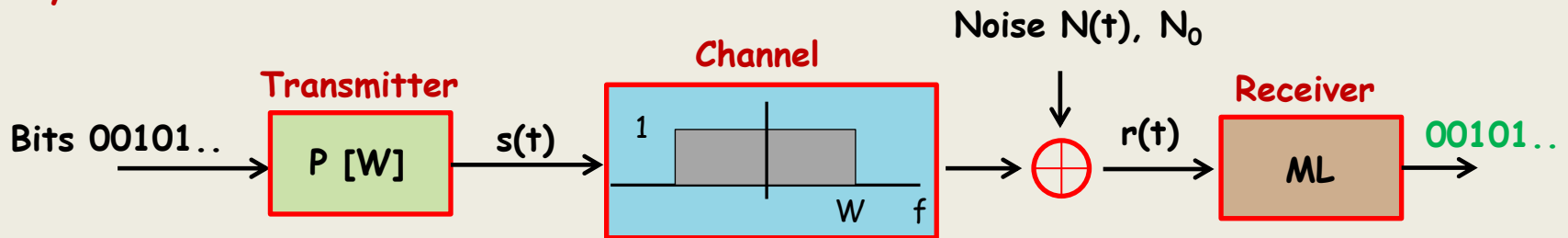
$$C_{\max} = \frac{P}{N_0 \ln(2)} = \frac{1}{\ln(2)}$$

**W**

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:

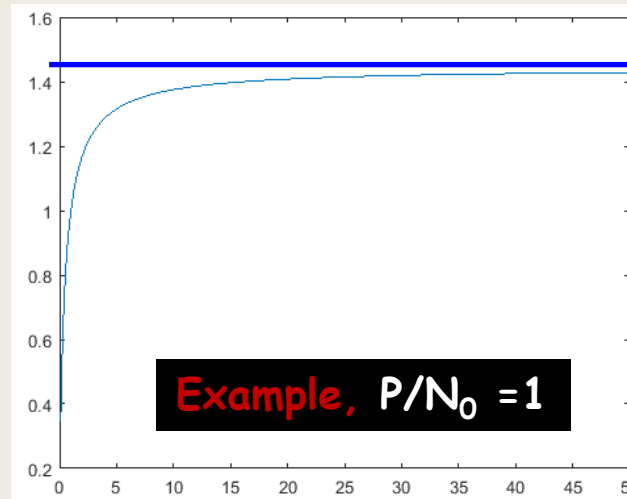


$$C = \frac{W}{\ln(2)} \ln \left( 1 + \frac{P}{N_0 W} \right)$$

Standard limit

$$\lim_{x \rightarrow \infty} x \ln \left( 1 + \frac{A}{x} \right) = A$$

**C**



But it grows to a limit

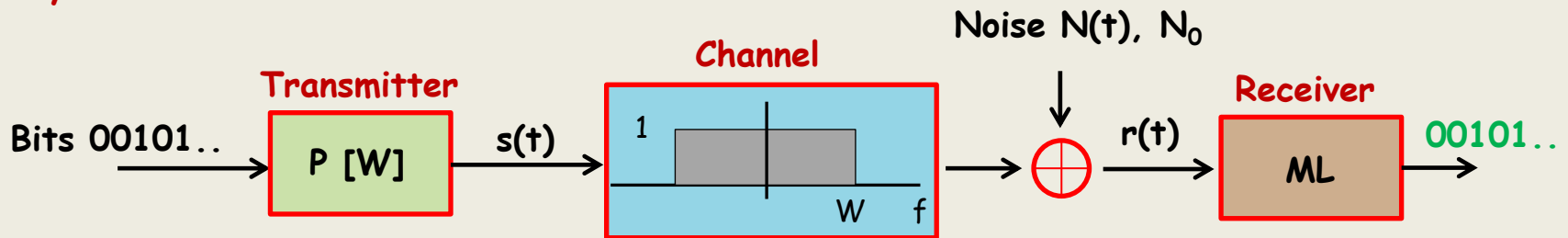
$$\begin{aligned} C_{\max} &= \frac{P}{N_0 \ln(2)} \\ &= \frac{1}{\ln(2)} \\ &= 1.4427 \end{aligned}$$

**W**

# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



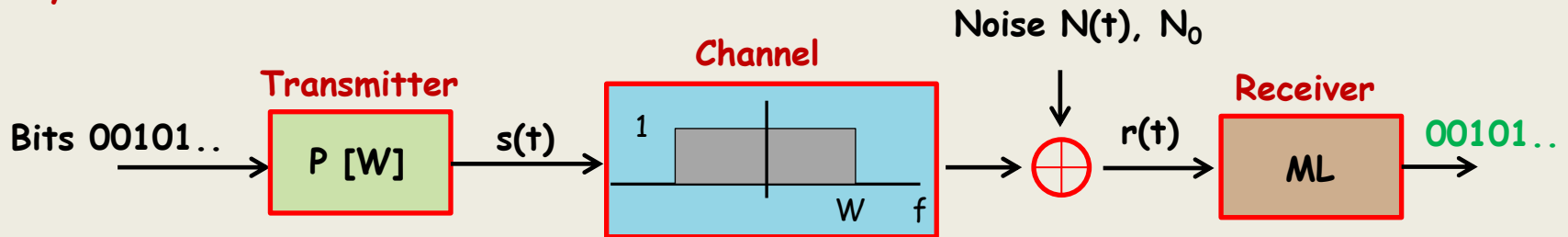
### Summary

1. We stated that the capacity of the above is  $C = W \log_2 \left( 1 + \frac{\mathcal{P}}{N_0 W} \right)$  bits/second
2. We proved that for infinite bandwidth, the capacity is  $C_{\max} = \frac{\mathcal{P}}{N_0 \ln(2)}$

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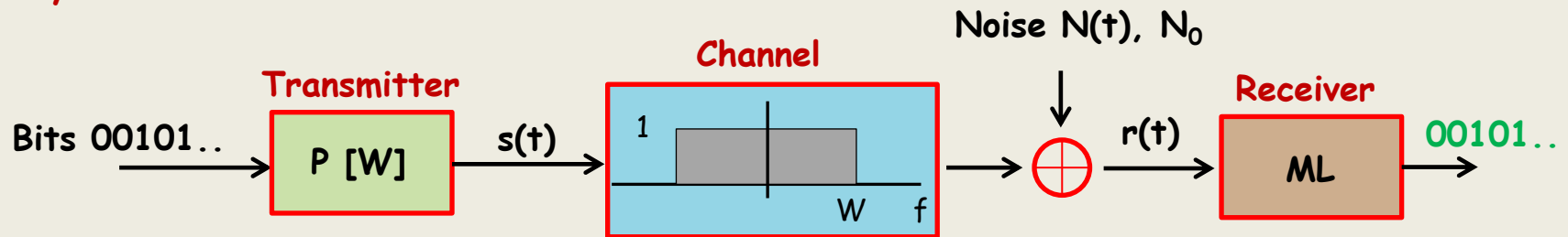
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# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



Bandwidth efficiency

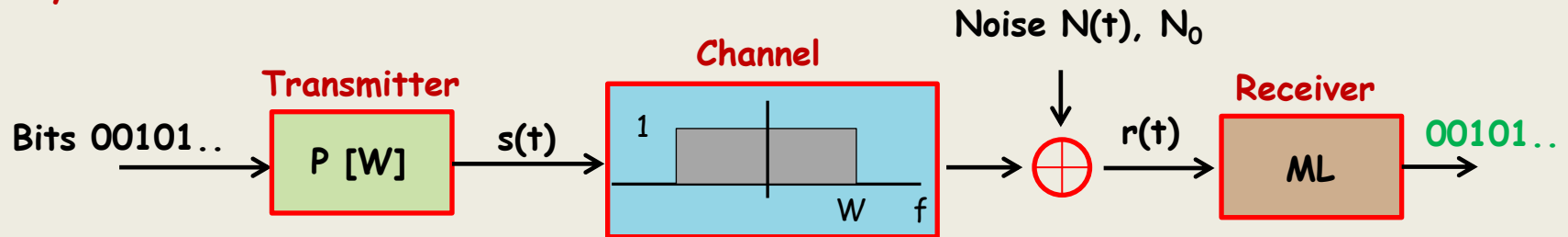
By definition,  $\frac{C}{W} = \log_2 \left( 1 + \frac{\mathcal{P}}{N_0 W} \right)$



# Lecture 4: Capacity

## Extension to continuous channel (Shannon '48)

System model:



Bandwidth efficiency

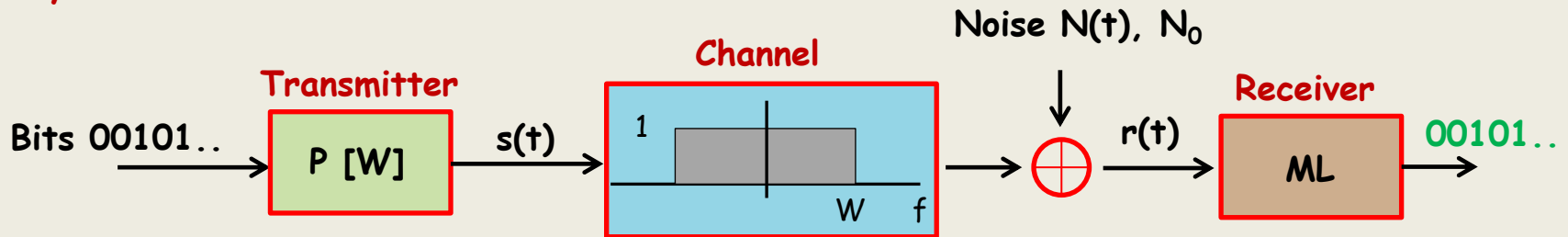
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Effect of increasing/decreasing  $W$  ?

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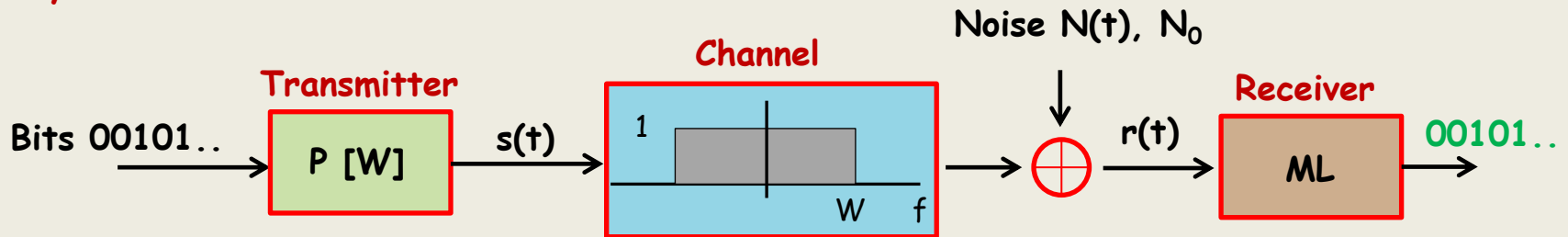
For large  $W$ , BW efficiency = 0

For small  $W$ , BW efficiency =  $\infty$

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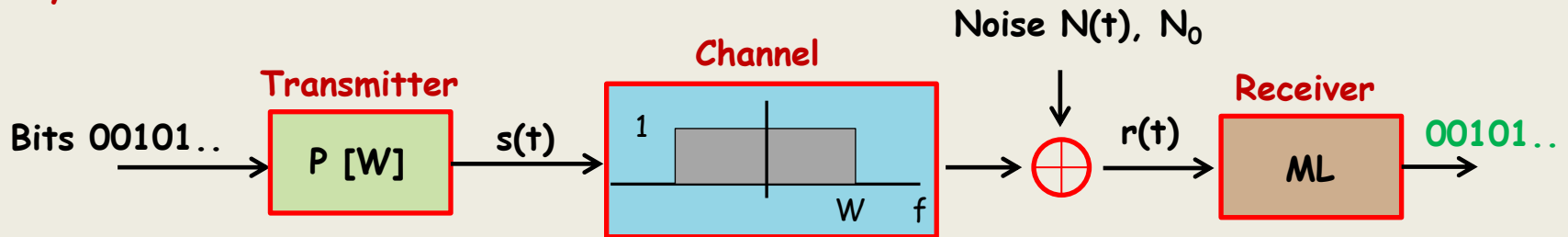
### Bandwidth vs. Power efficiency

However  $P = C E_b$

# Lecture 4: Capacity

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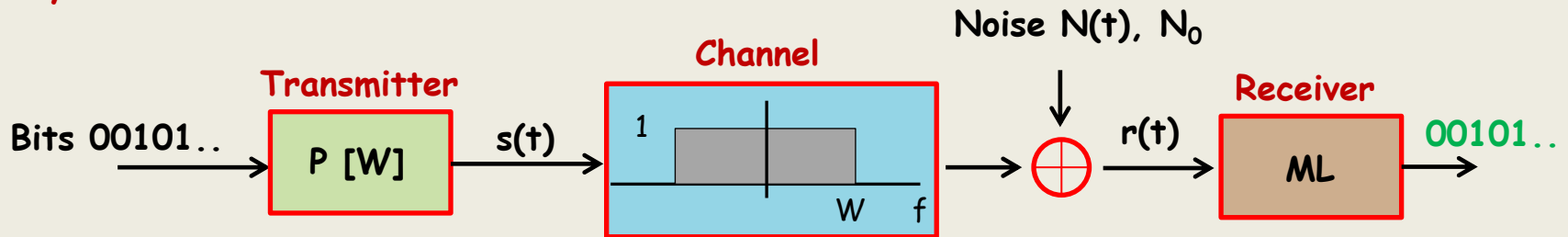
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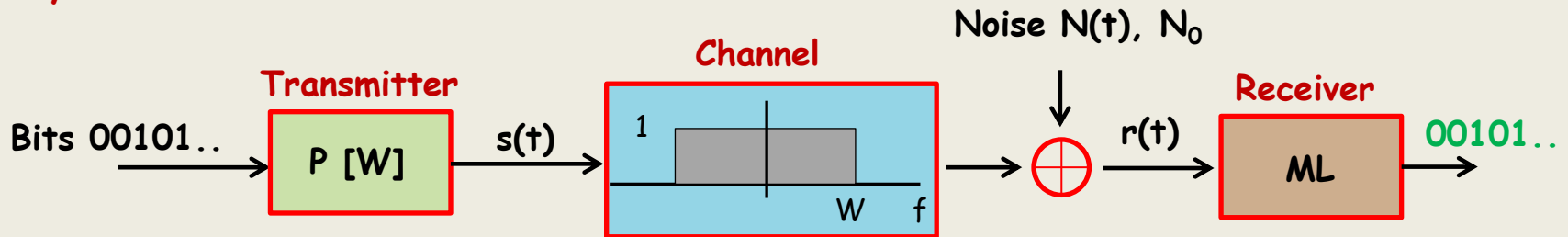
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Or, equivalently  $\frac{E_b}{N_0} = \frac{2^{\frac{C}{W}} - 1}{\frac{C}{W}}$

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## Extension to continuous channel (Shannon '48)

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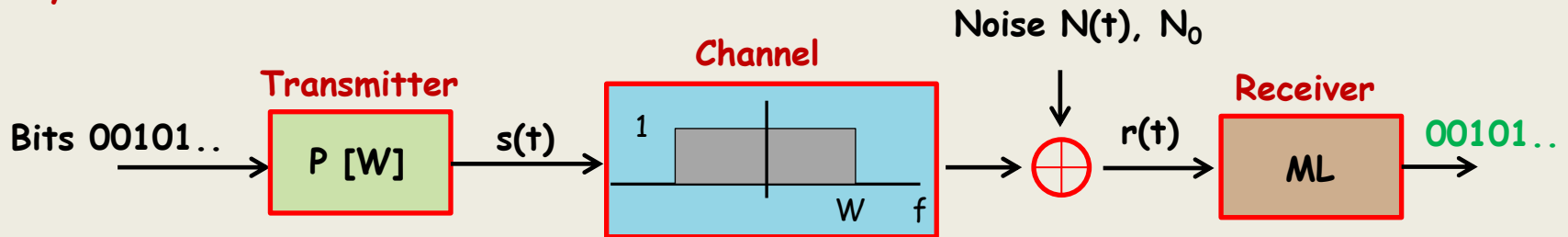
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### Bandwidth vs. Power efficiency

What happens if  $C/W$  grows?

$E_b/N_0$  grows as well

In fact, we have (check at home) that to have 0 error probability

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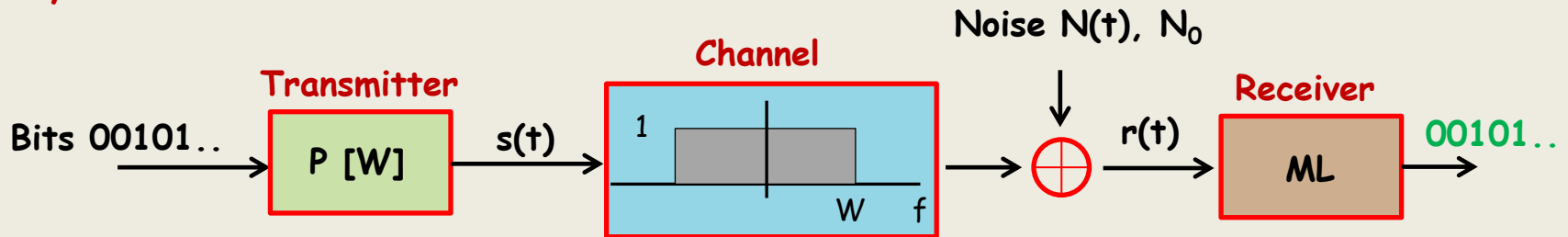
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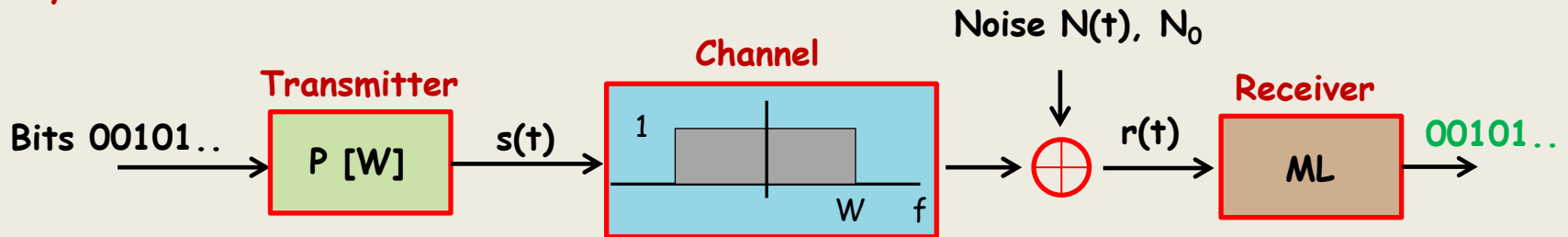
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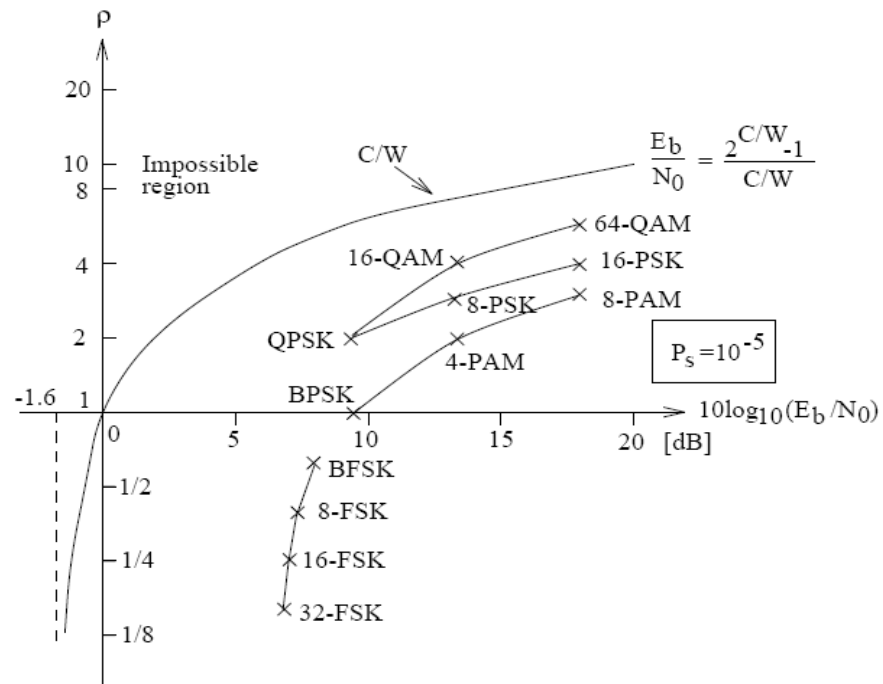
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Thus

$\lim_{C/W \rightarrow 0} \frac{E_b}{N_0} \geq \ln(2) \quad (= -1.6\text{dB})$

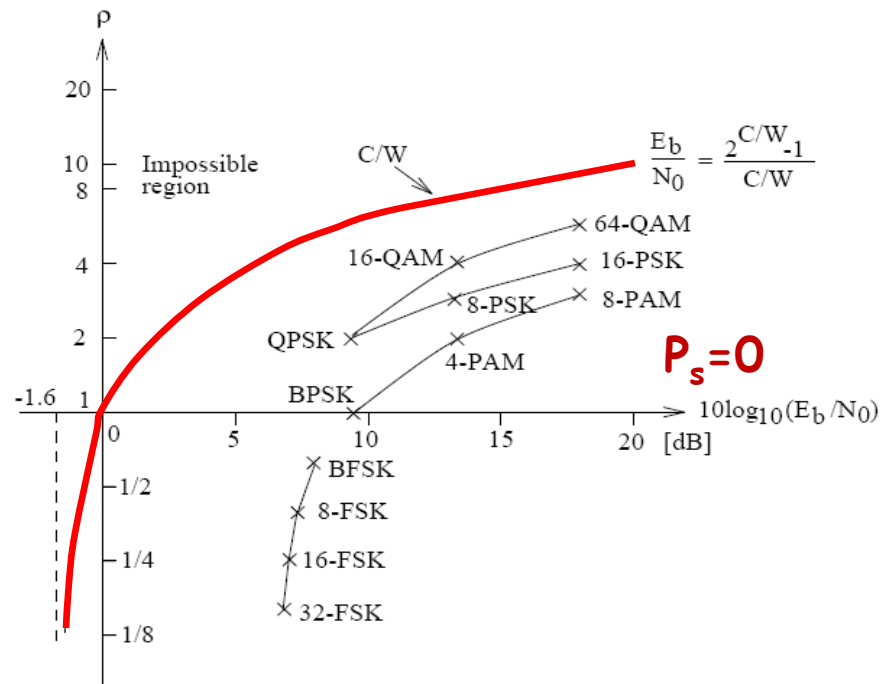
# Lecture 4: Capacity



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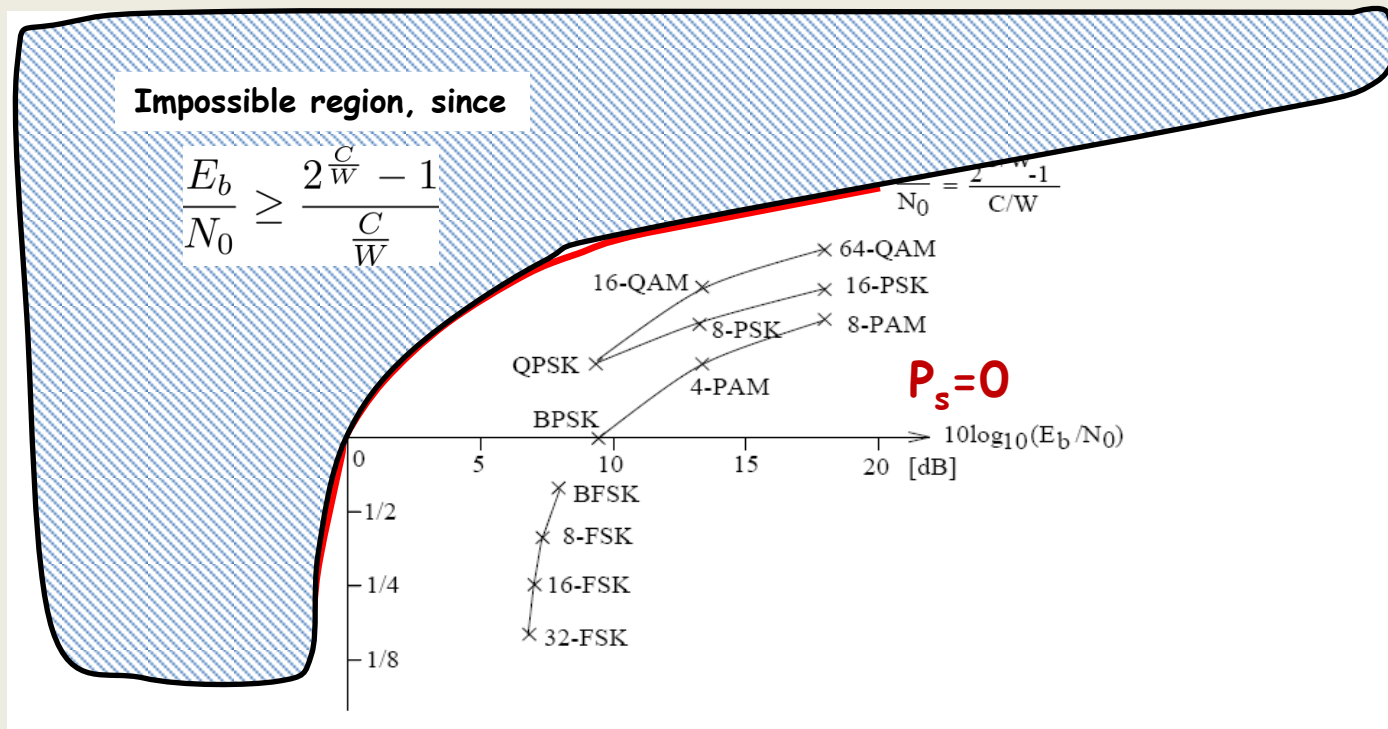
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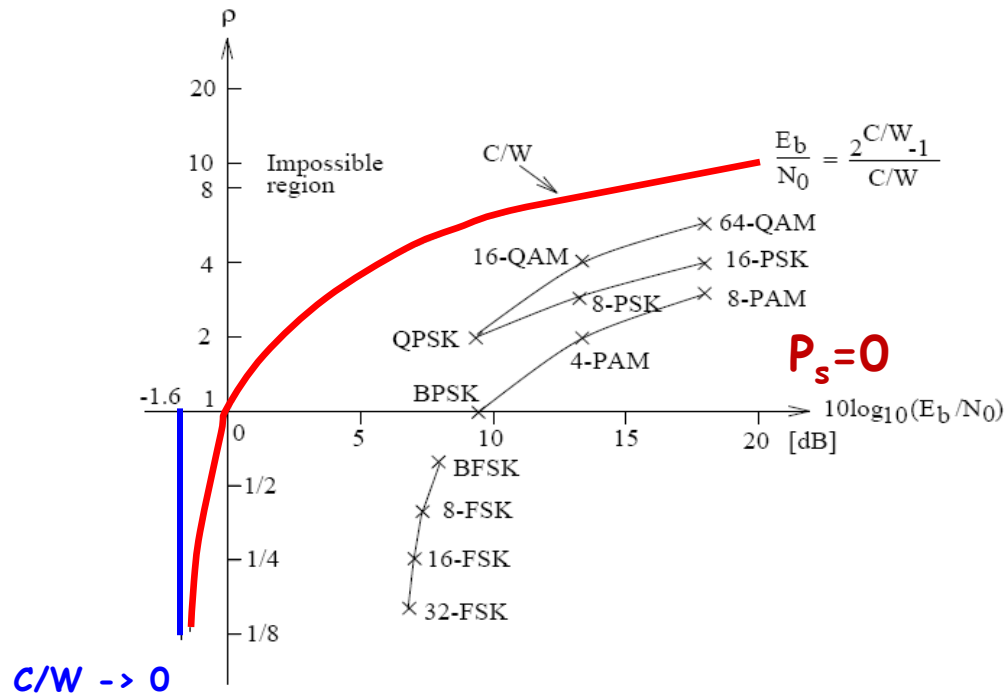


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NOTE: Plot does not tell what the capacity is in bit/sec

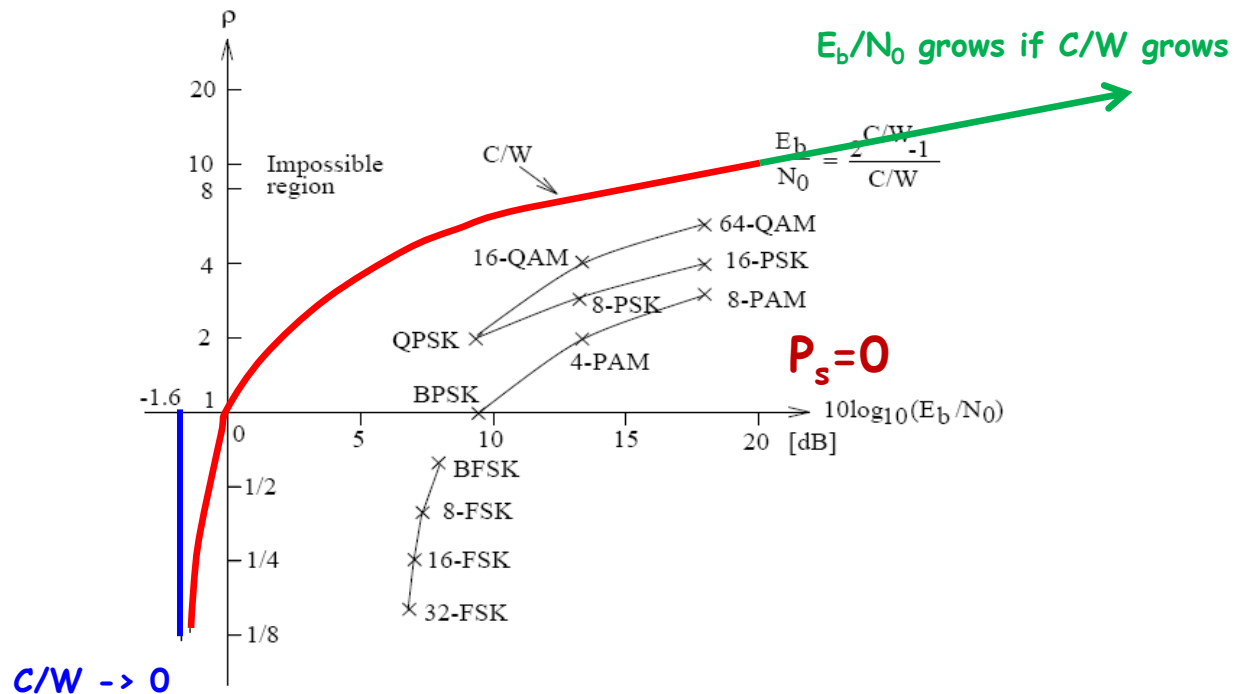
# Lecture 4: Capacity



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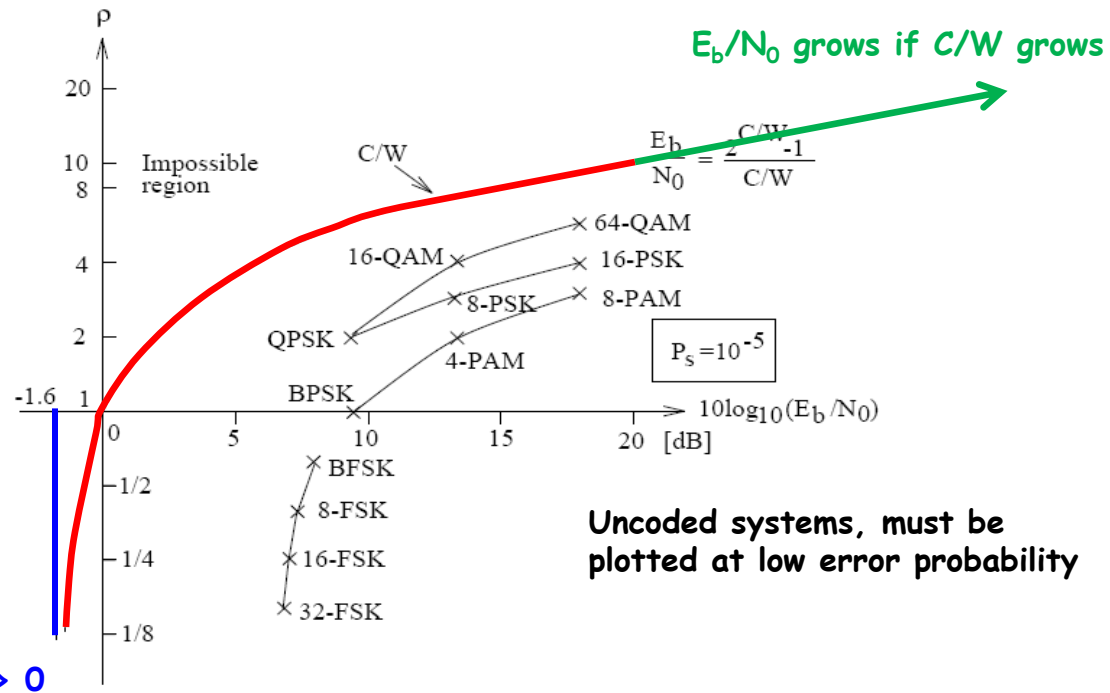
# Lecture 4: Capacity



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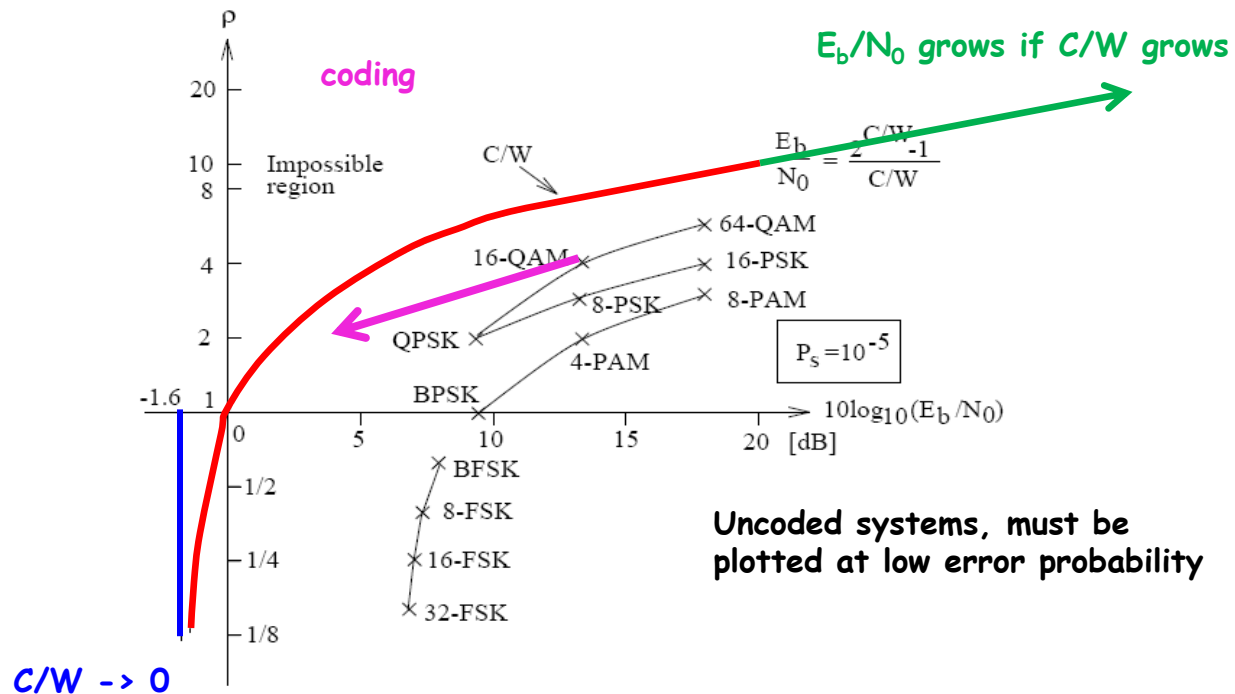
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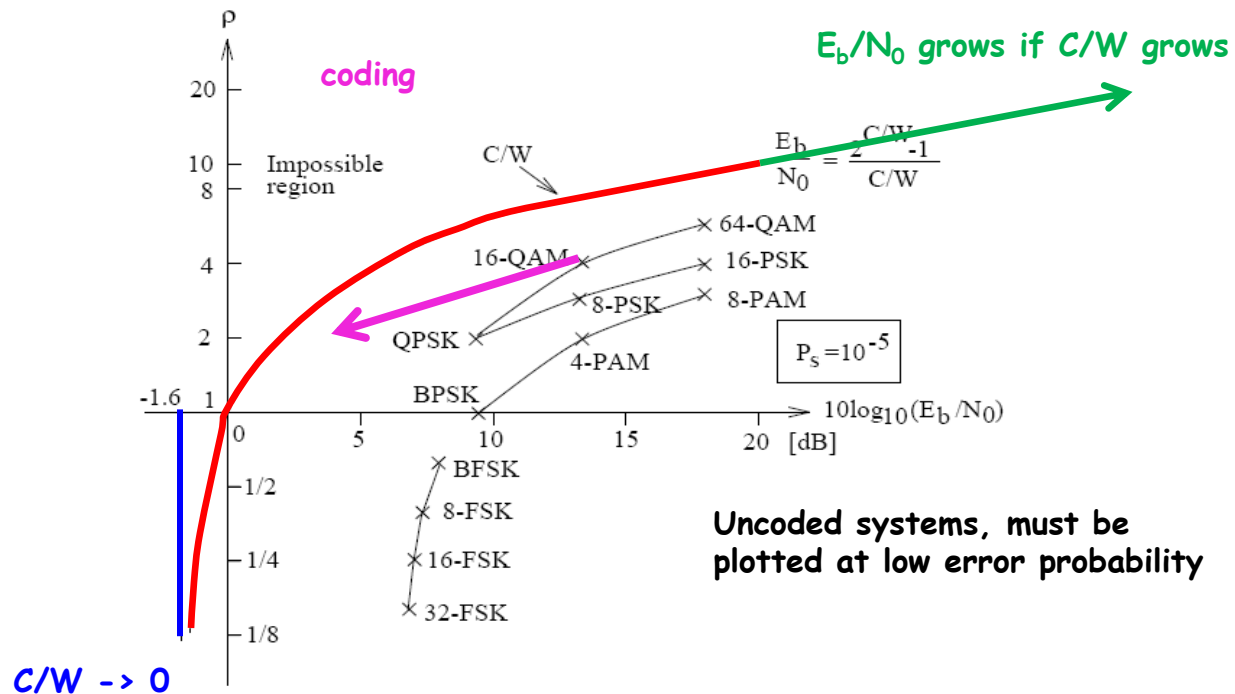


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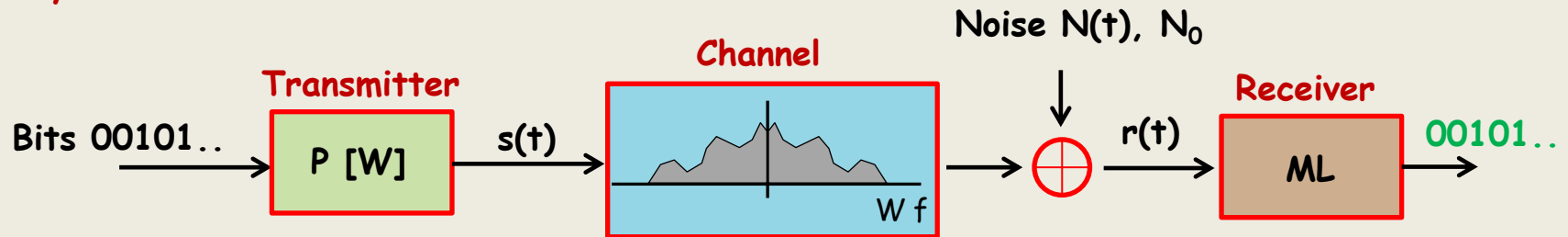
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# Lecture 4: Capacity

## Extension to frequency dependent channel

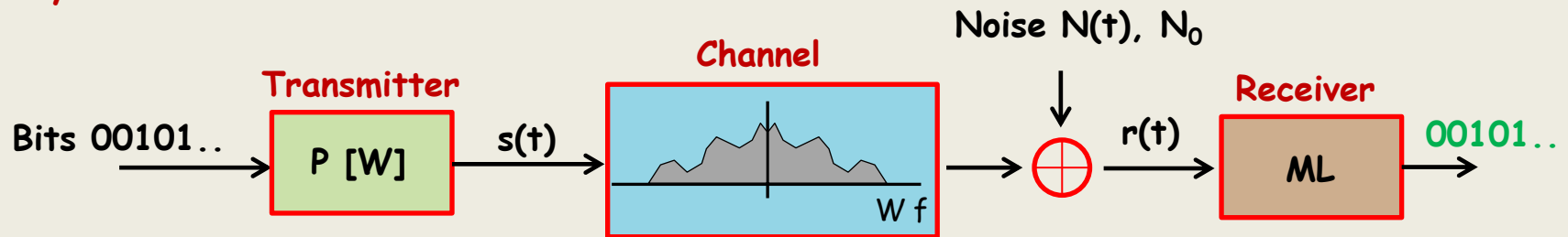
System model:



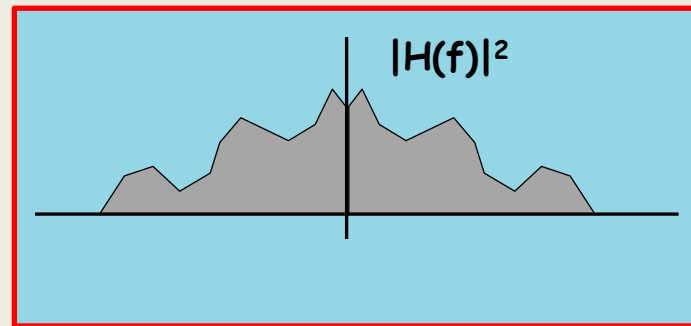
# Lecture 4: Capacity

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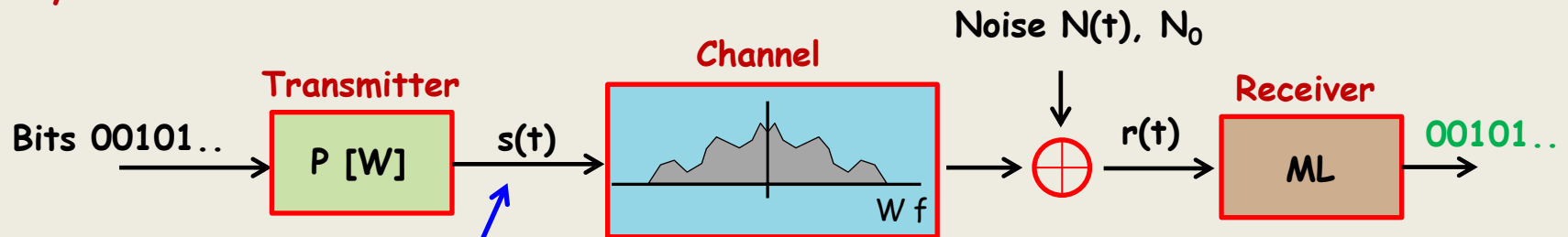
Frequency response of channel



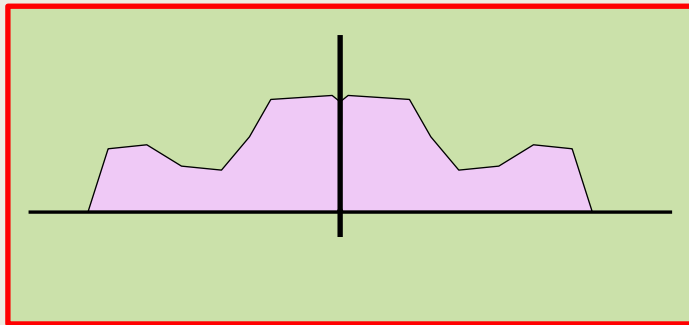
# Lecture 4: Capacity

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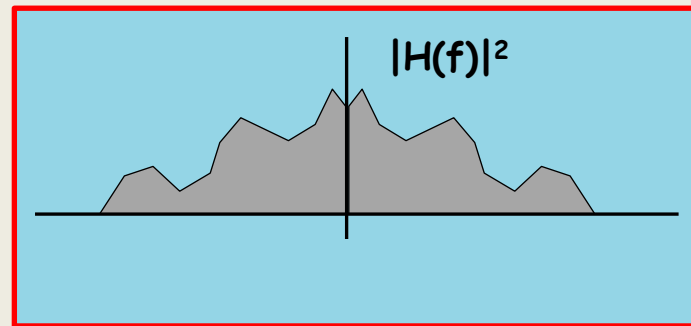
System model:



Signal also have  
frequency representation



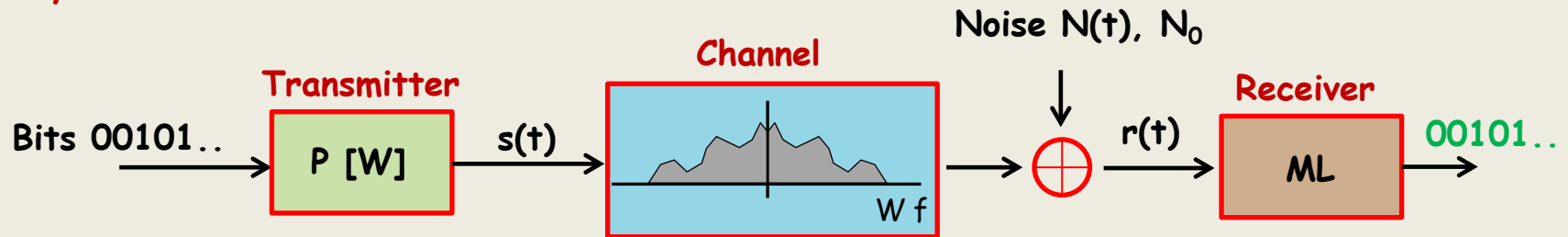
Frequency response of channel



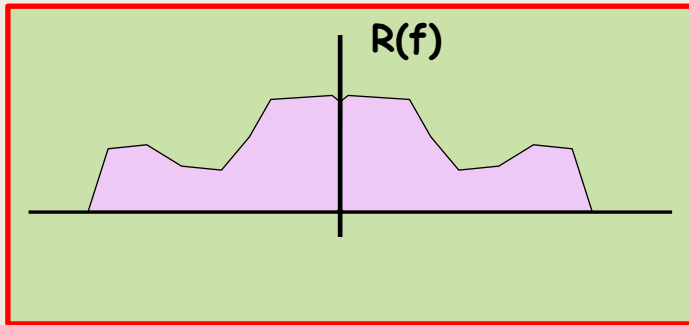
# Lecture 4: Capacity

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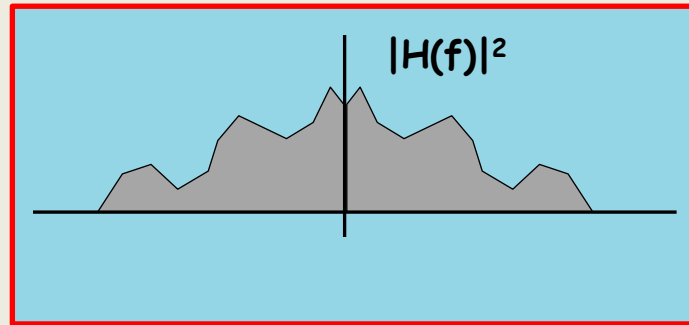
System model:



Power spectral density is what matters



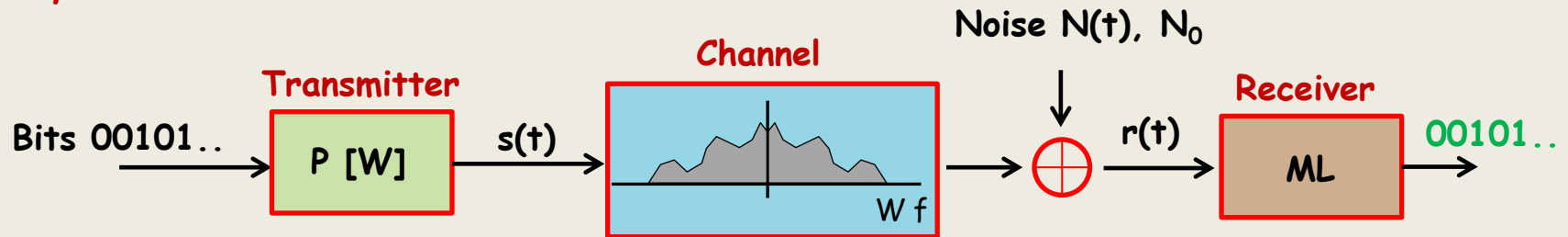
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# Lecture 4: Capacity

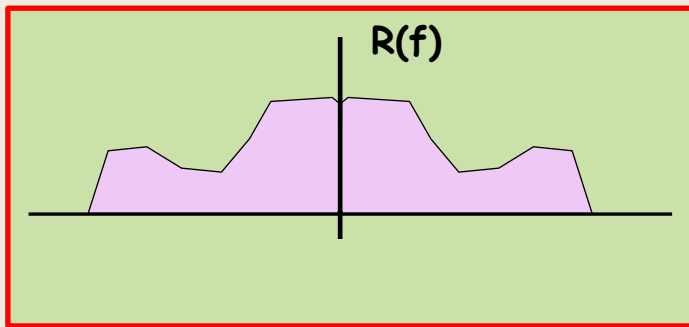
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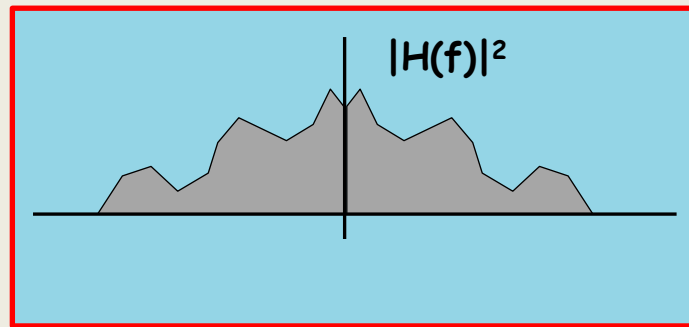


Constraint on  $R(f)$  ?

Power spectral density is what matters



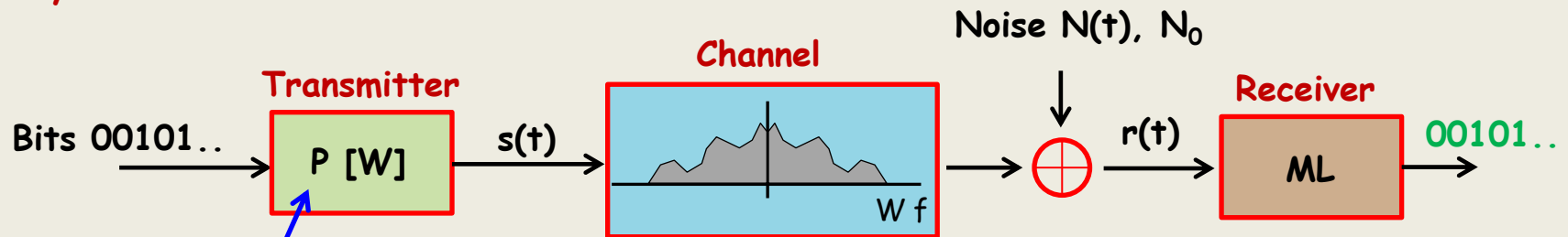
Frequency response of channel



# Lecture 4: Capacity

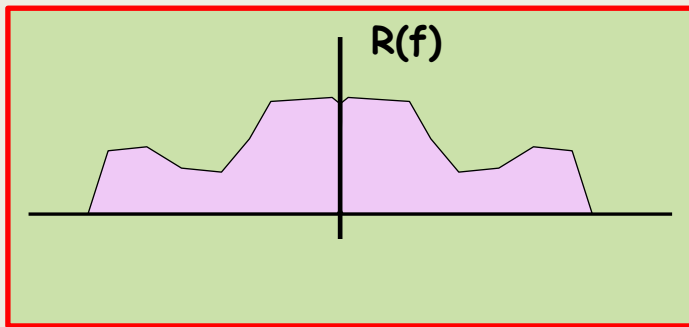
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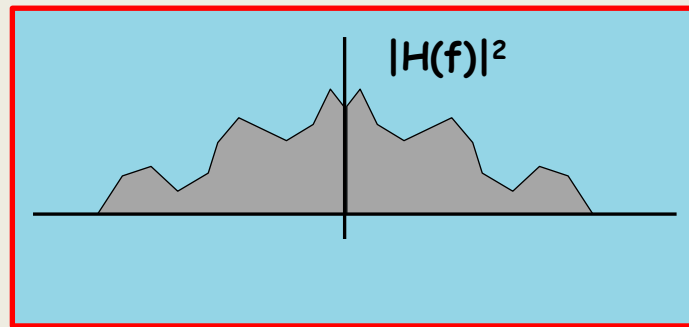


Constraint on  $R(f)$ ?  $\int_{-\infty}^{\infty} R(f) df = P$

Power spectral density is what matters



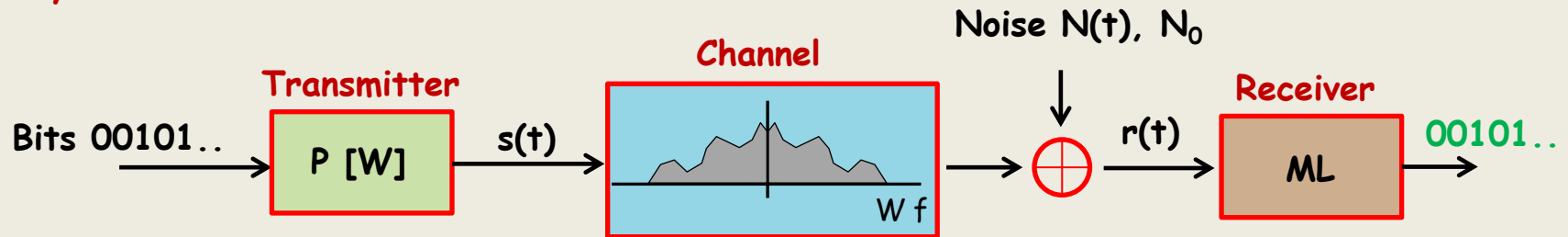
Frequency response of channel



# Lecture 4: Capacity

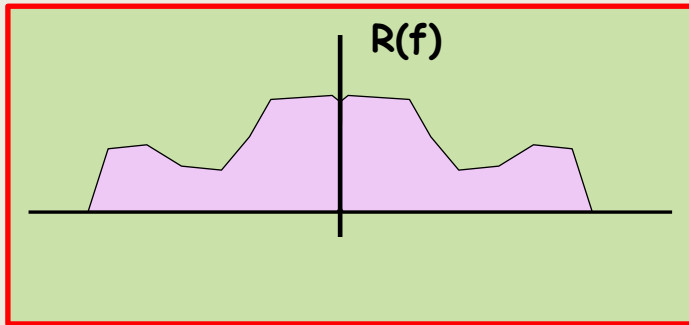
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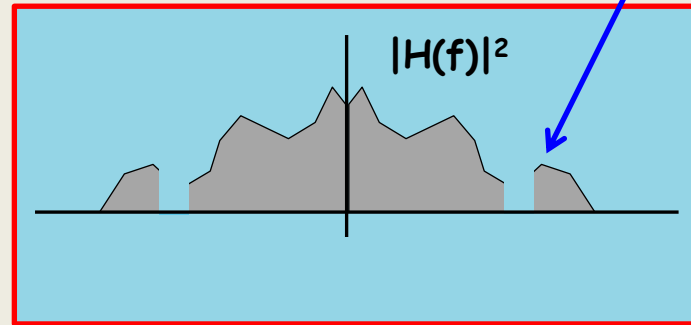
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If deep fade here

Frequency response of channel

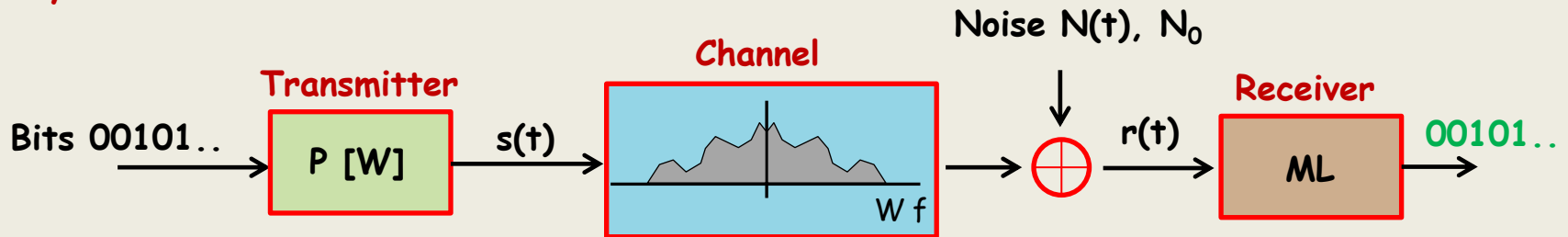




# Lecture 4: Capacity

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System model:

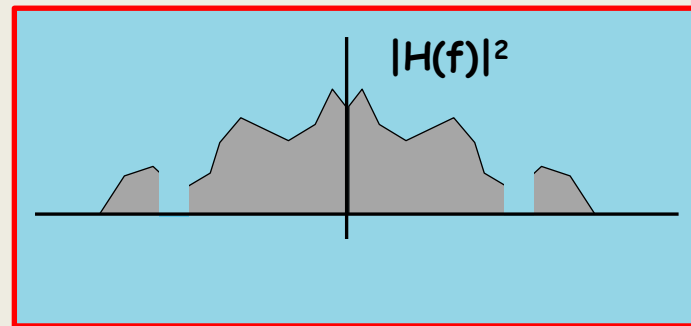
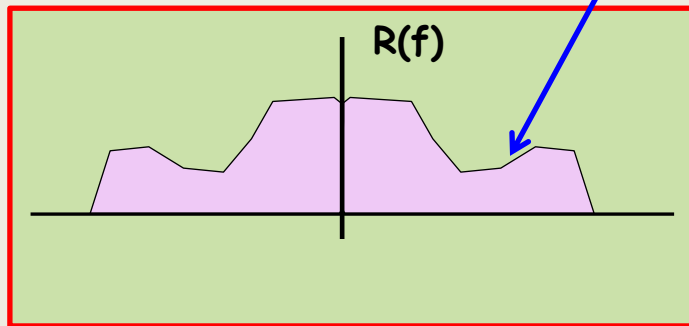


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Power spectral density is what matters

Stupidity to put power here

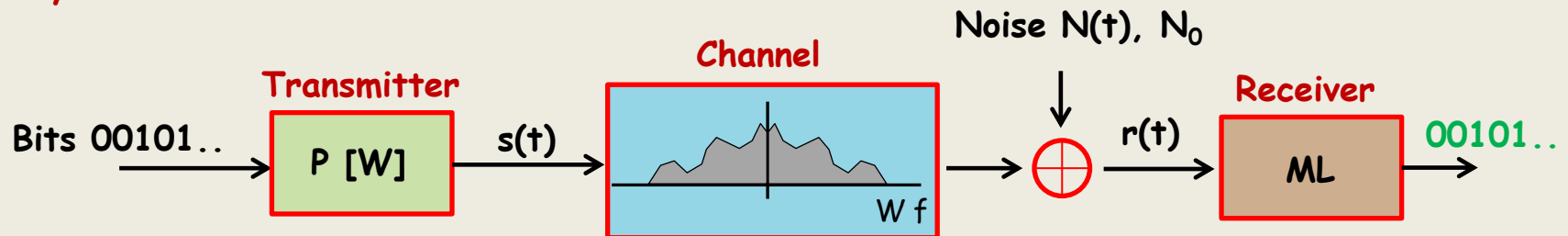
Frequency response of channel



# Lecture 4: Capacity

## Extension to frequency dependent channel

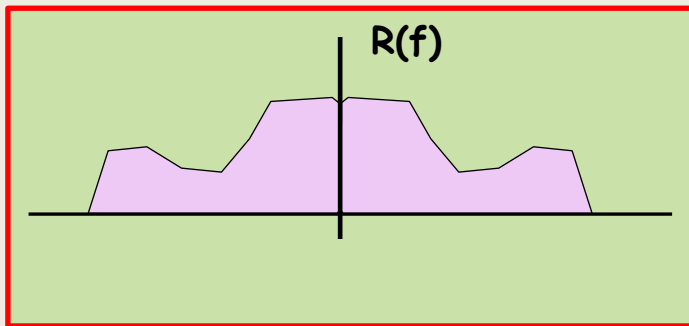
System model:



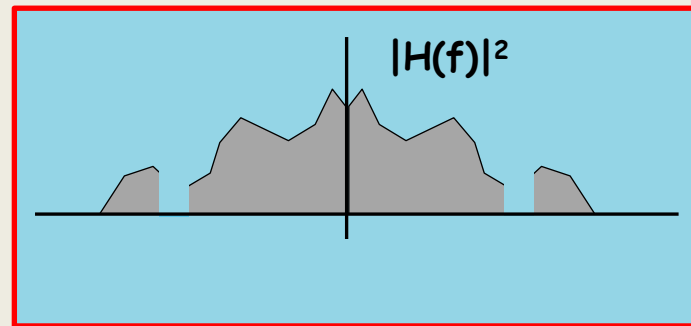
Conclusion: We should optimize the left plot, for the given right plot

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Power spectral density is what matters

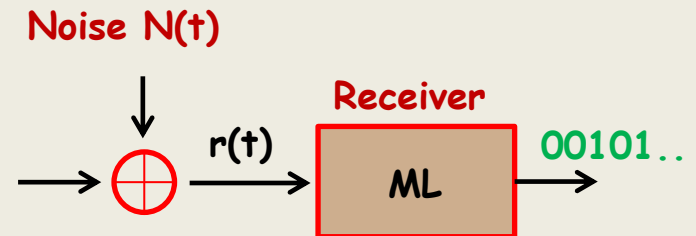
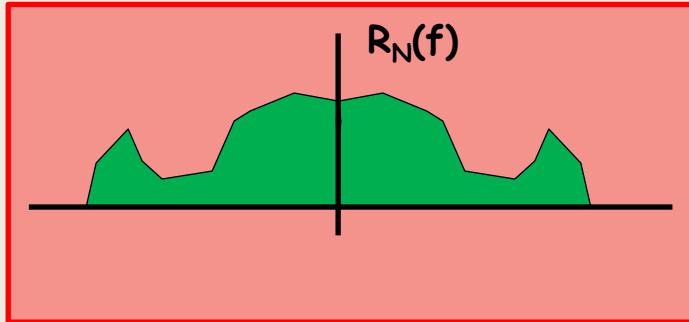


Frequency response of channel

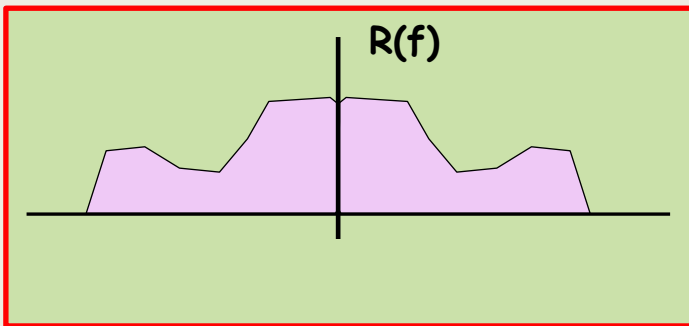


# Lecture 4: Capacity

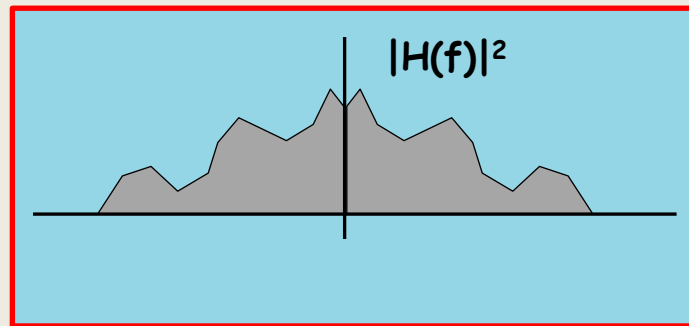
Frequency response of Noise



Power spectral density is what matters

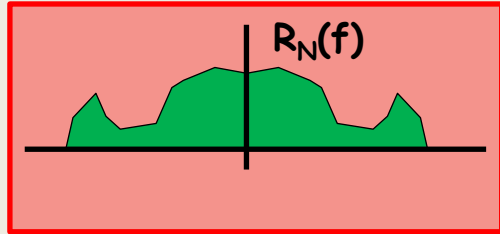


Frequency response of channel

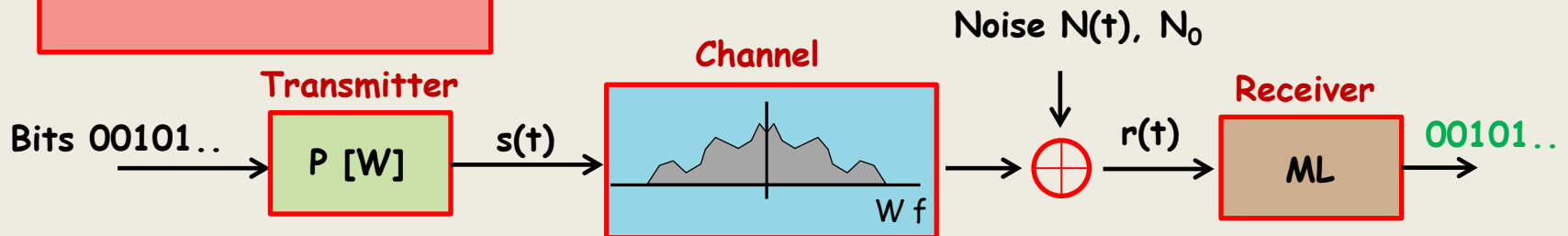


# Lecture 4: Capacity

## Frequency response of Noise



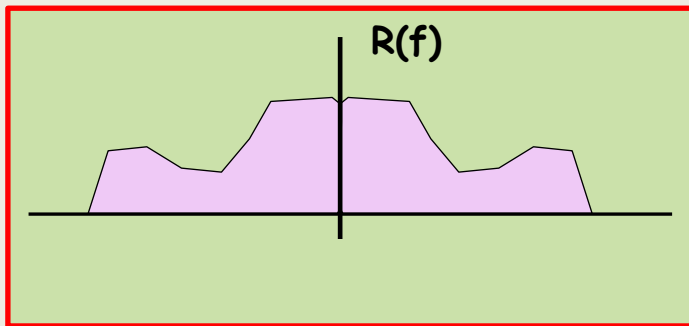
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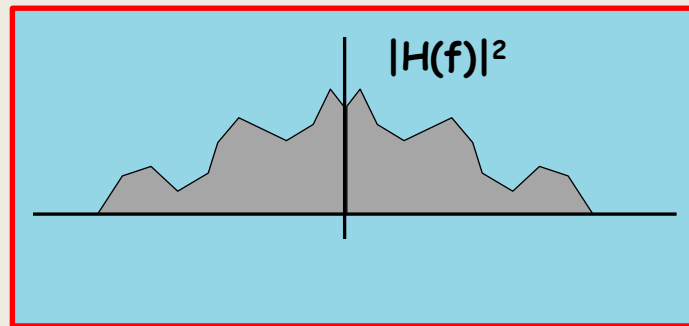
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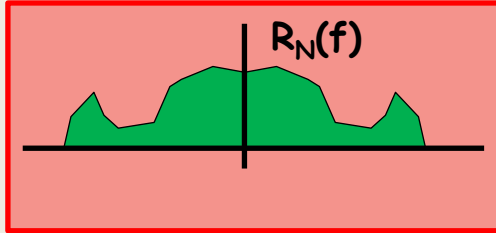


Frequency response of channel



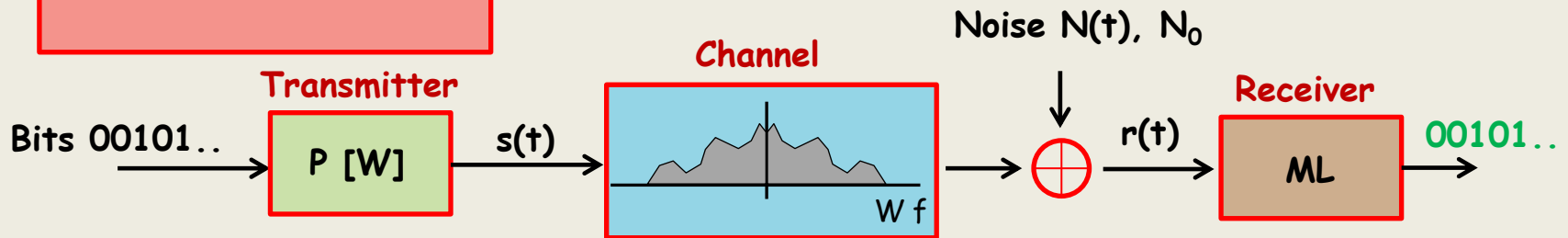
# Lecture 4: Capacity

Frequency response of Noise



We need to find this formula

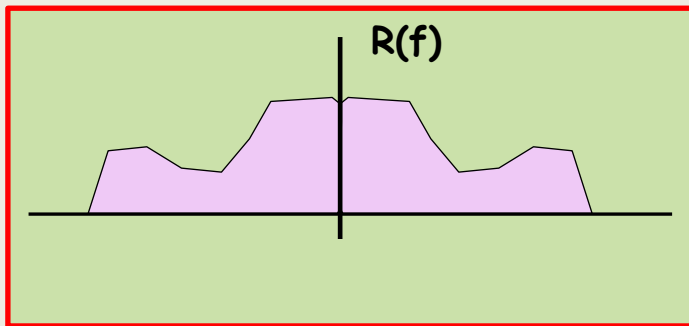
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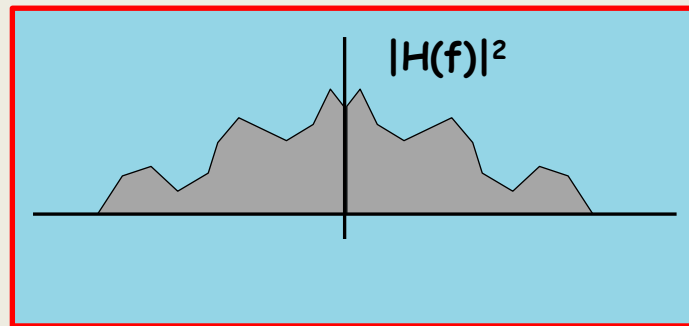
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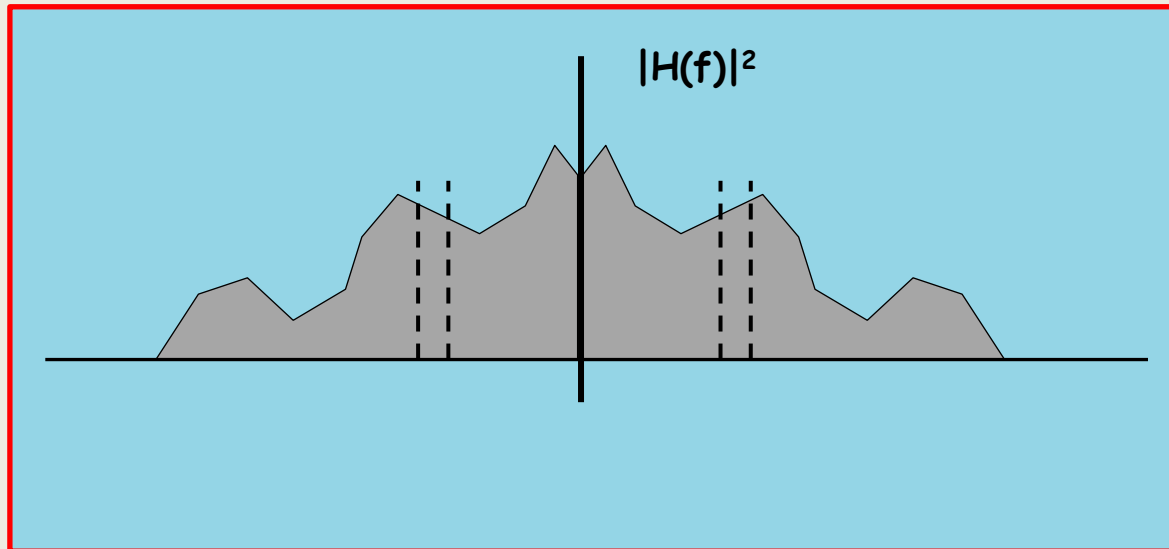


Frequency response of channel



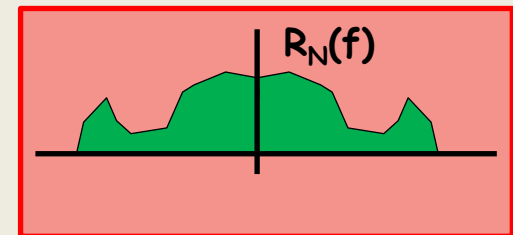
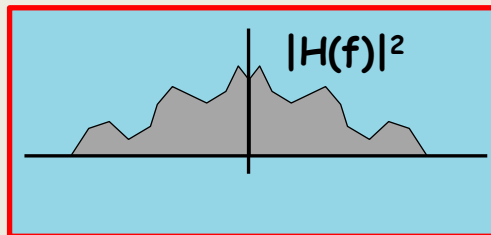
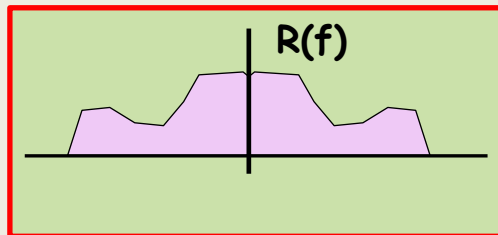
# Lecture 4: Capacity

$$C = \max_{R(f): \int R(f) df = P} \text{Capacity}(|H(f)|^2, R_N(f), R(f))$$



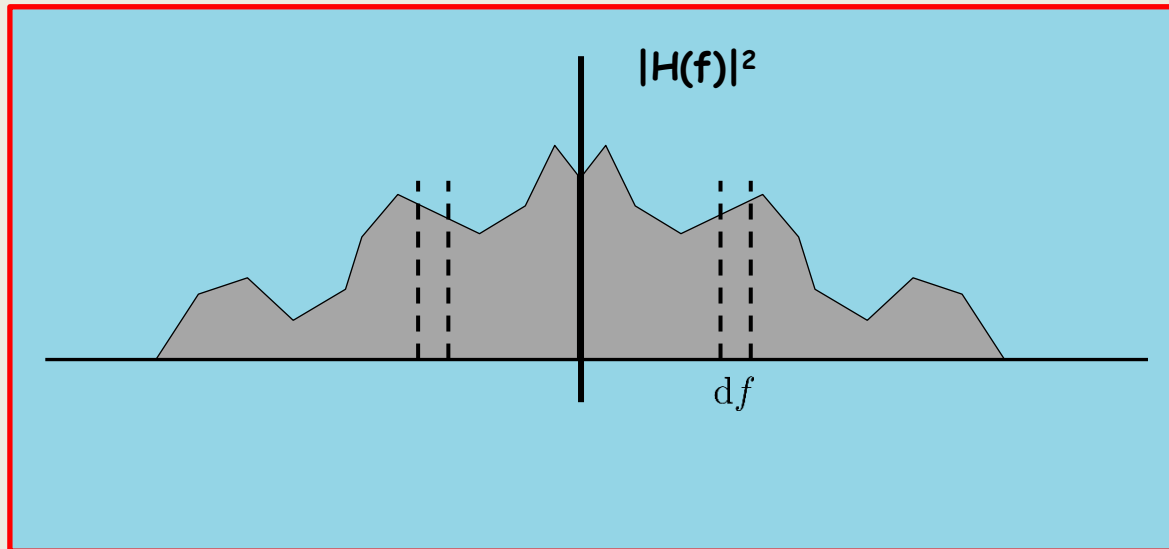
In this small piece  
We can use

$$C = W \log_2 \left( 1 + \frac{P}{N_0 W} \right)$$



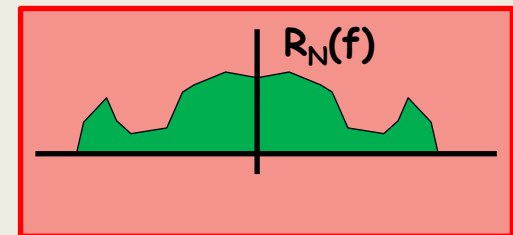
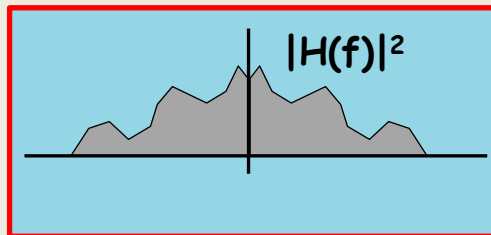
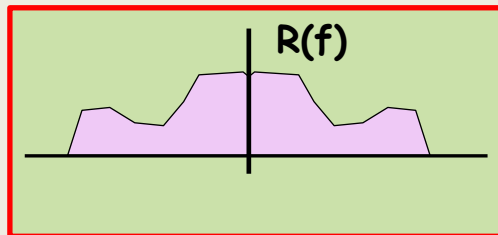
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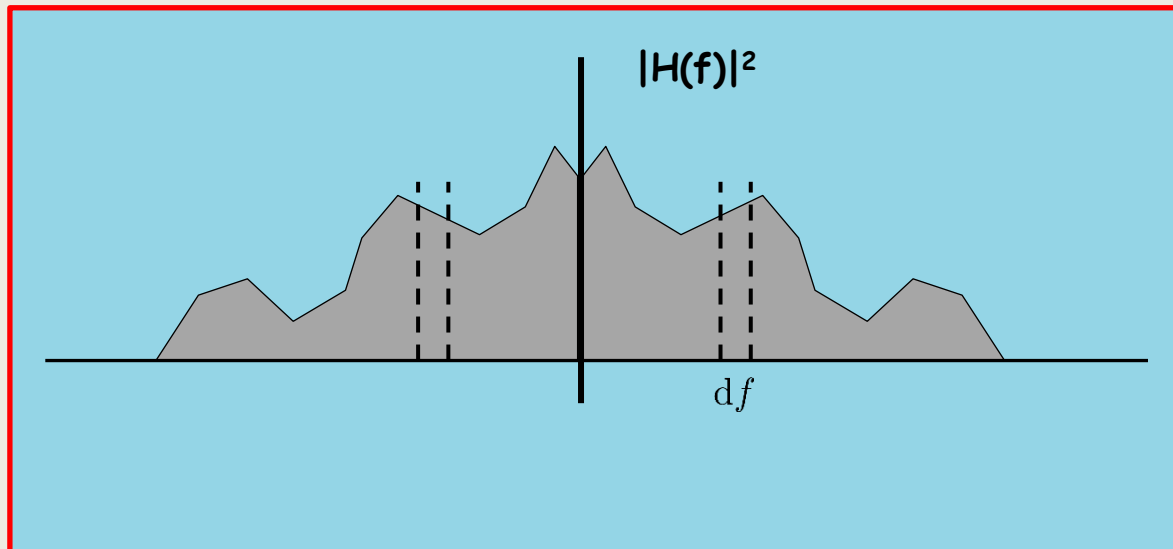
In this small piece  
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$$C = df \log_2 \left( 1 + \frac{P}{N_0 df} \right)$$



# Lecture 4: Capacity

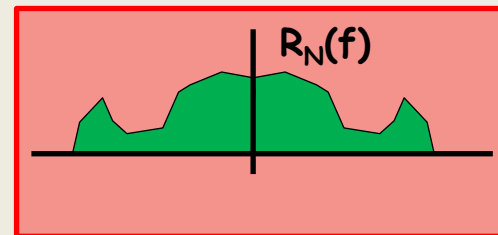
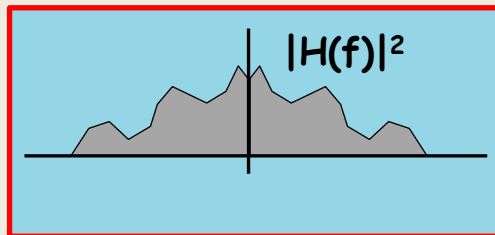
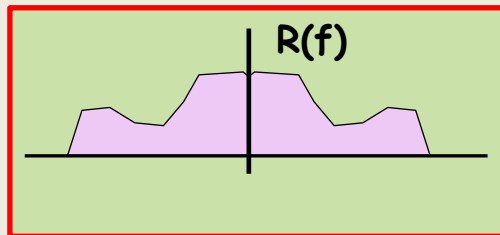
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In this small piece  
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$$C = df \log_2 \left( 1 + \frac{P}{N_0 df} \right)$$

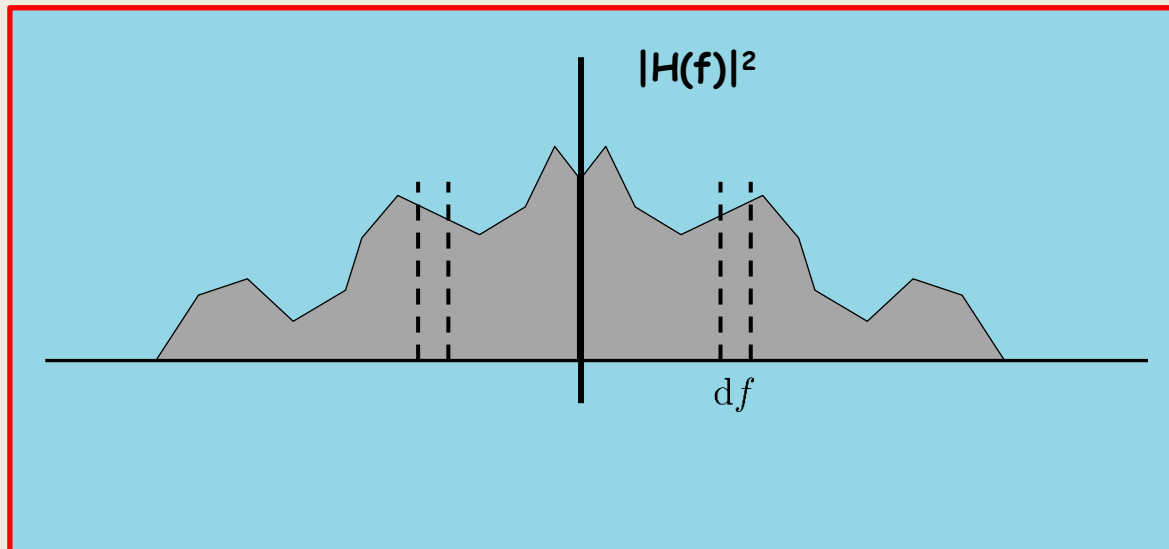
For flat noise,  
 $R_N(f) = N_0/2$





# Lecture 4: Capacity

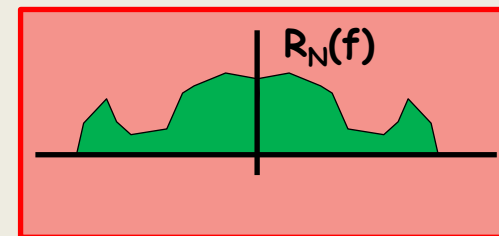
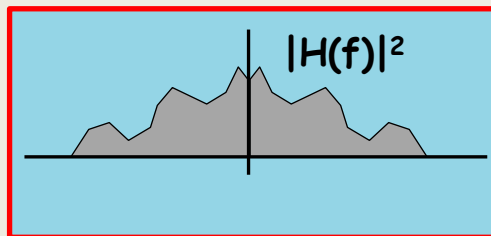
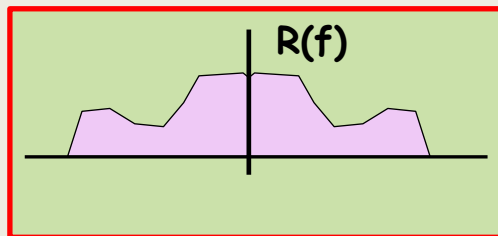
$$C = \max_{R(f): \int R(f) df = P} \text{Capacity}(|H(f)|^2, R_N(f), R(f))$$



In this small piece  
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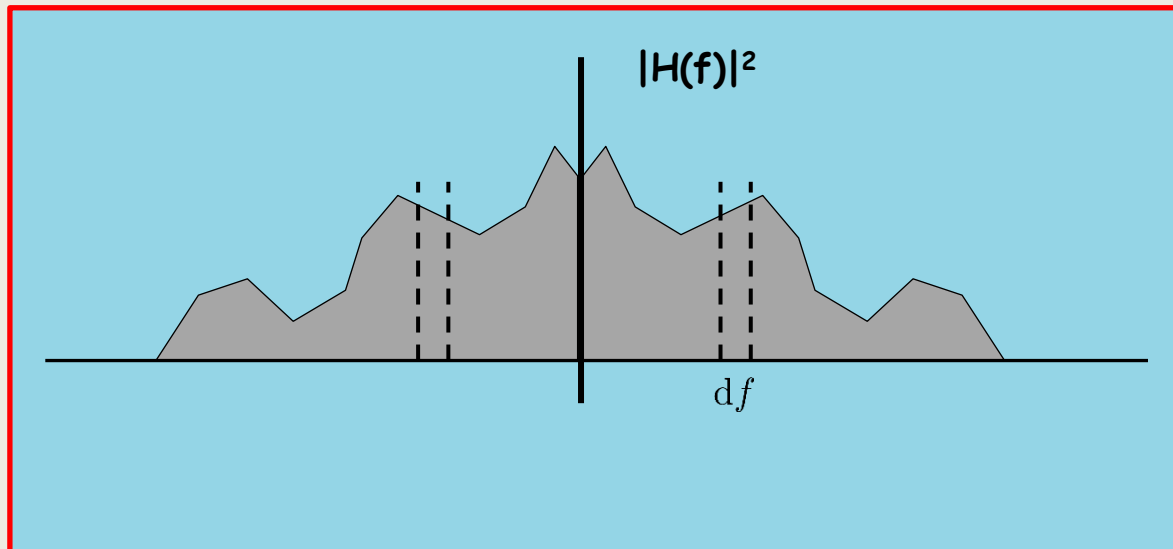
$$C = df \log_2 \left( 1 + \frac{P}{N_0 df} \right) = 2R_N(f)$$

For flat noise,  
 $R_N(f) = N_0/2$



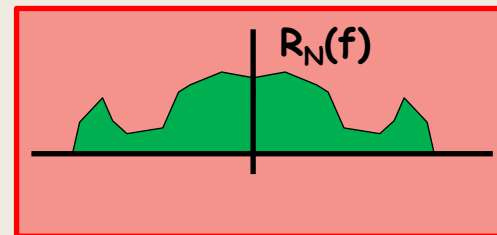
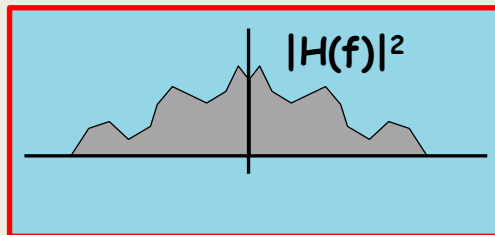
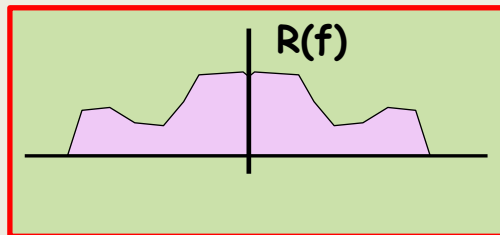
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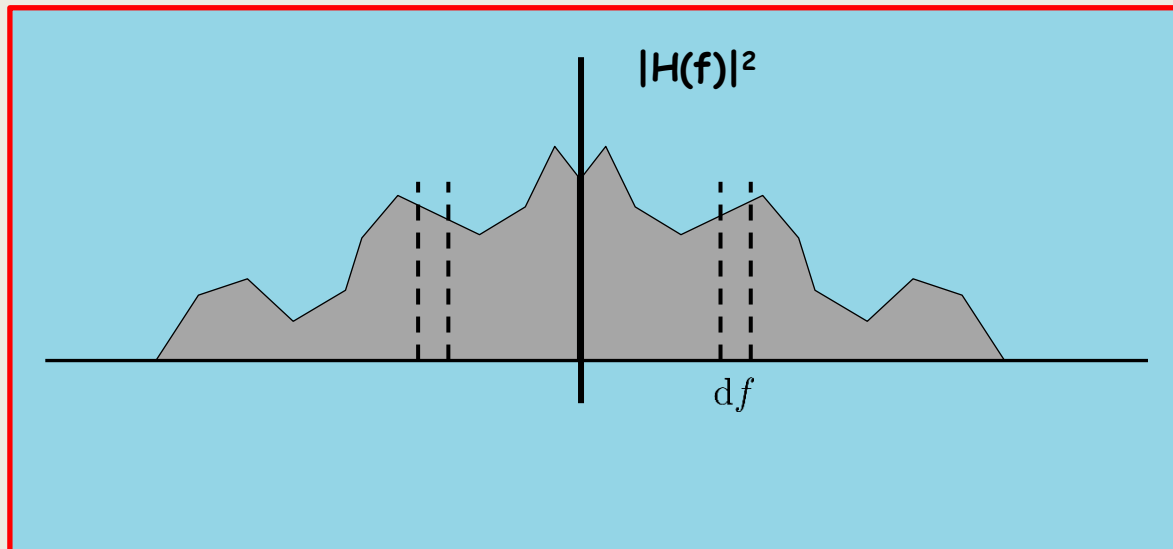
In this small piece  
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$$C = df \log_2 \left( 1 + \frac{P}{2R_N(f)df} \right)$$



# Lecture 4: Capacity

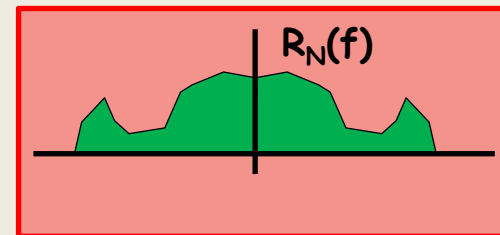
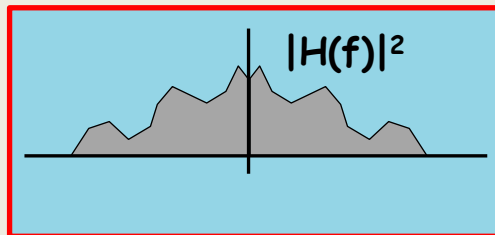
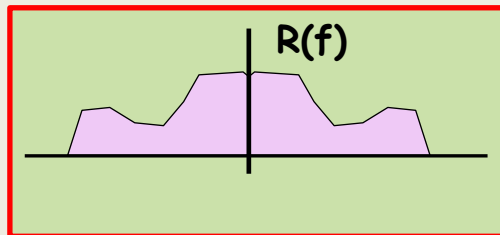
$$C = \max_{R(f): \int R(f) df = P} \text{Capacity}(|H(f)|^2, R_N(f), R(f))$$



In this small piece  
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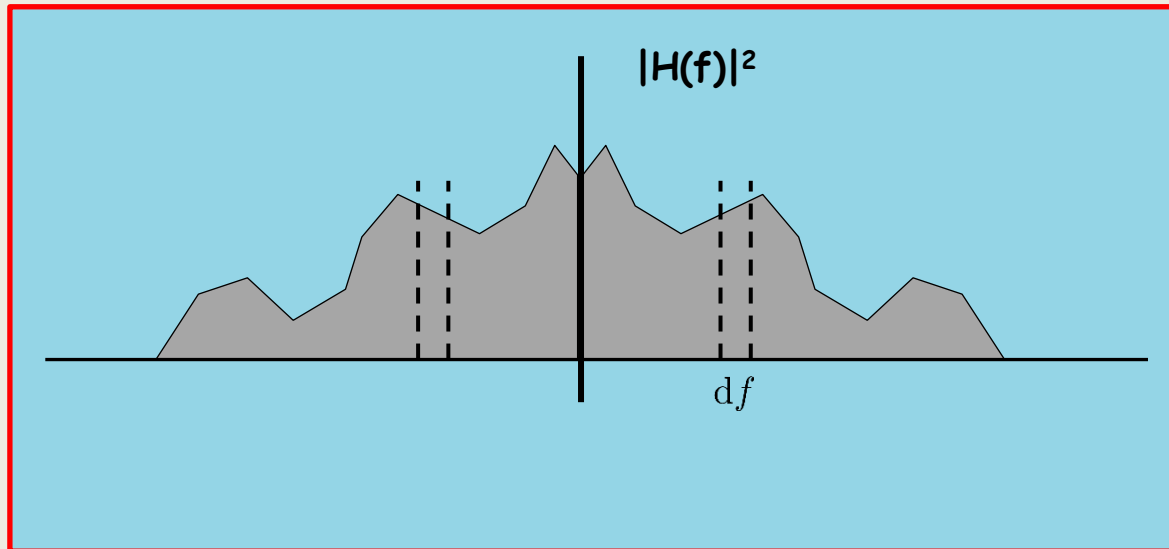
$$C = df \log_2 \left( 1 + \frac{P}{2R_N(f)df} \right)$$

How much power do  
we have?



# Lecture 4: Capacity

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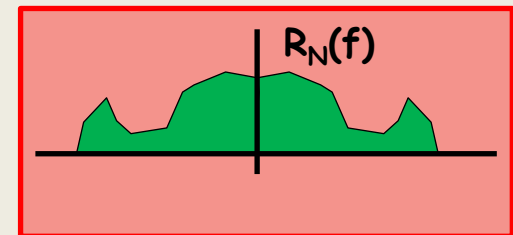
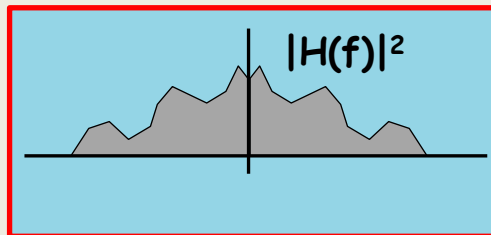
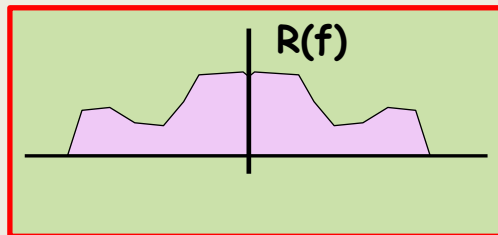


In this small piece  
We can use

$$C = df \log_2 \left( 1 + \frac{P}{2R_N(f)df} \right)$$

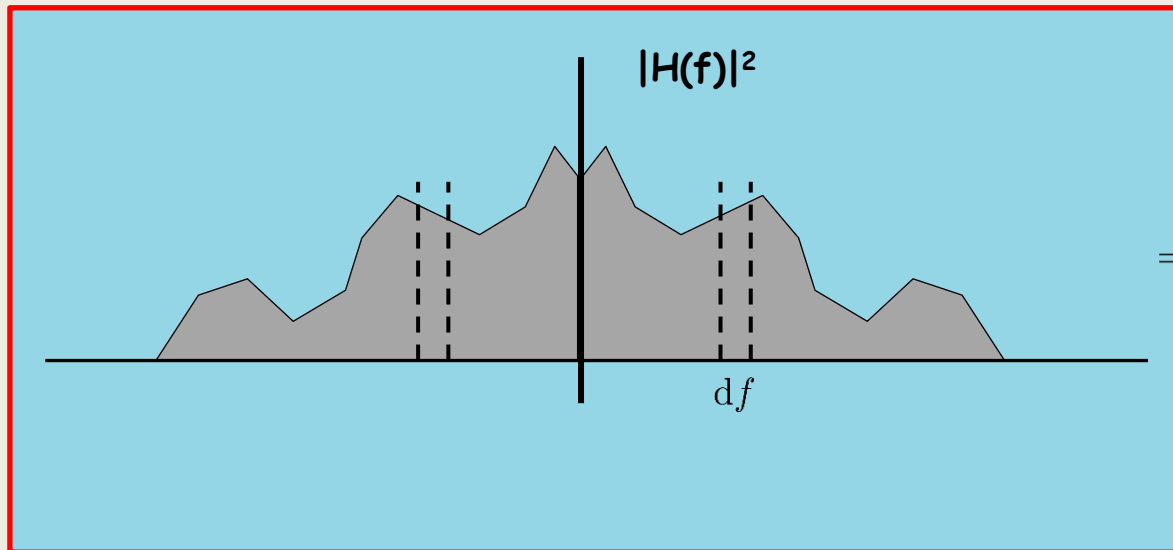
How much power do  
we have?

$$2df R(f) |H(f)|^2$$



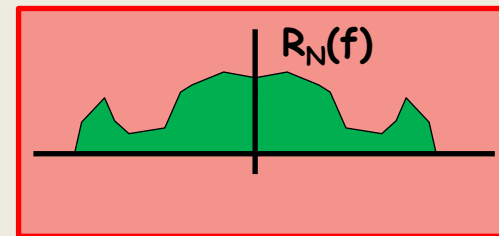
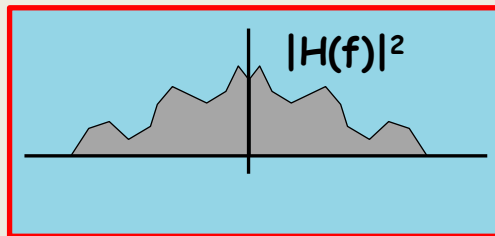
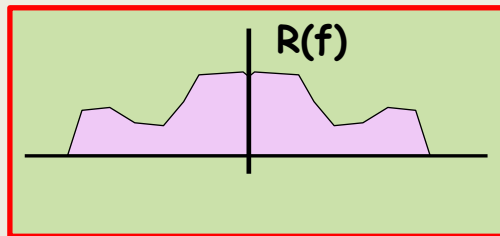
# Lecture 4: Capacity

$$C = \max_{R(f): \int R(f) df = P} \text{Capacity}(|H(f)|^2, R_N(f), R(f))$$



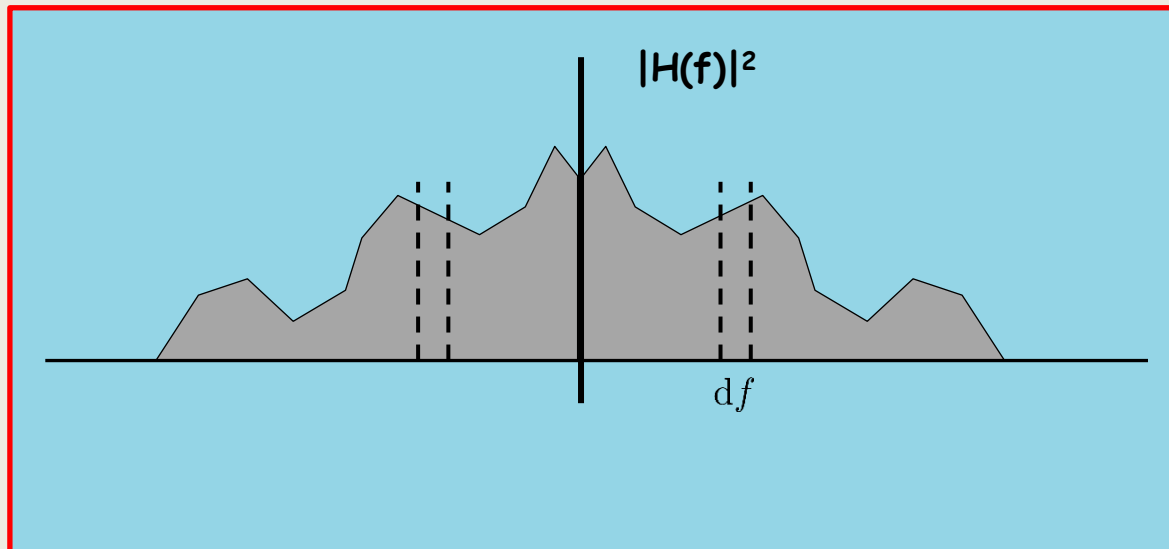
In this small piece  
We can use

$$C = df \log_2 \left( 1 + \frac{P}{2R_N(f)df} \right) \\ = df \log_2 \left( 1 + \frac{2R(f)|H(f)|^2 df}{2R_N(f)df} \right)$$



# Lecture 4: Capacity

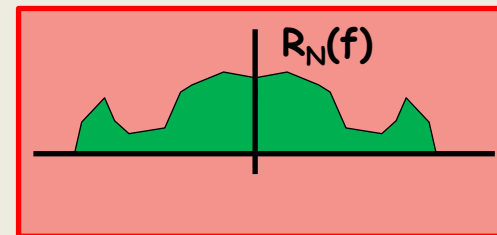
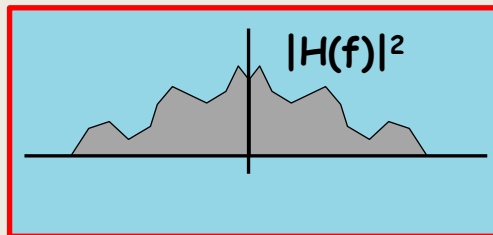
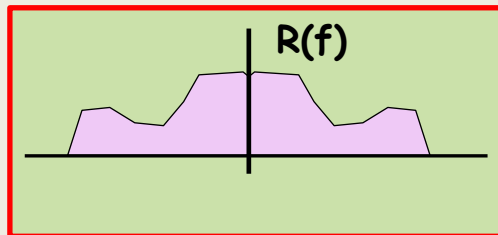
$$C = \max_{R(f): \int R(f) df = P} \text{Capacity}(|H(f)|^2, R_N(f), R(f))$$



In this small piece  
We can use

$$C = df \log_2 \left( 1 + \frac{P}{2R_N(f)df} \right)$$

$$= df \log_2 \left( 1 + \frac{R(f)|H(f)|^2}{R_N(f)} \right)$$



# Lecture 4: Capacity

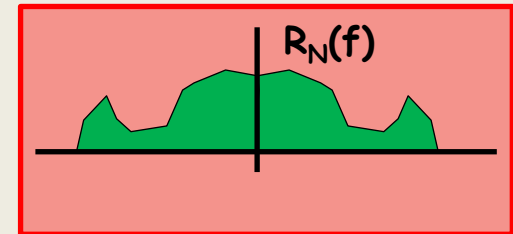
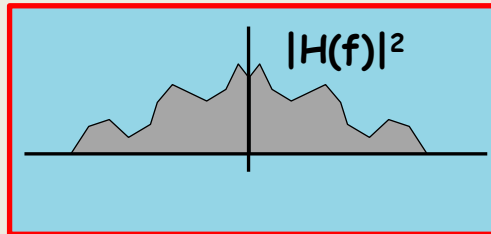
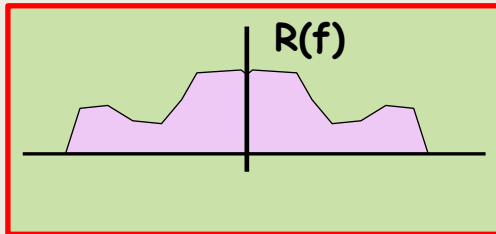
$$C = \max_{R(f): \int R(f) df = P} \text{Capacity}(|H(f)|^2, R_N(f), R(f))$$

Sum up

$$\text{Capacity}(|H(f)|^2, R_N(f), R(f)) = \int_0^\infty \log_2 \left( 1 + \frac{R(f)|H(f)|^2}{R_N(f)} \right) df$$

In this small piece  
We can use

$$\begin{aligned} C &= df \log_2 \left( 1 + \frac{P}{2R_N(f)df} \right) \\ &= df \log_2 \left( 1 + \frac{R(f)|H(f)|^2}{R_N(f)} \right) \end{aligned}$$

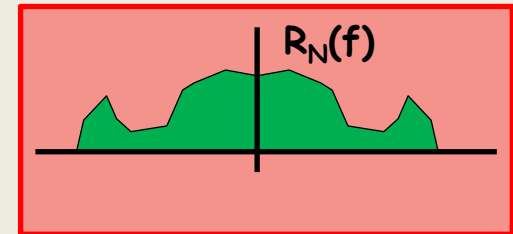
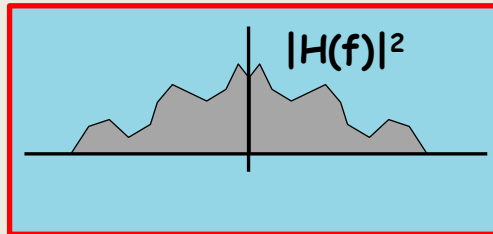
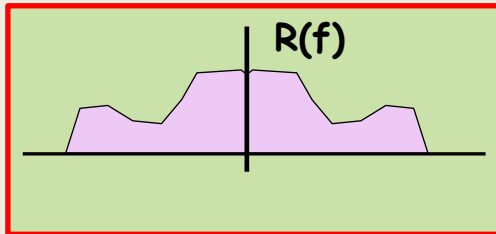


# Lecture 4: Capacity

$$C = \max_{R(f): \int R(f) df = P} \text{Capacity}(|H(f)|^2, R_N(f), R(f))$$

Sum up

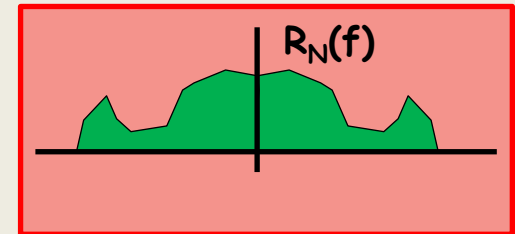
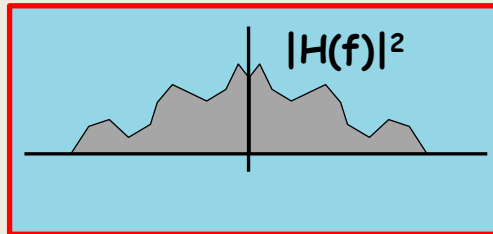
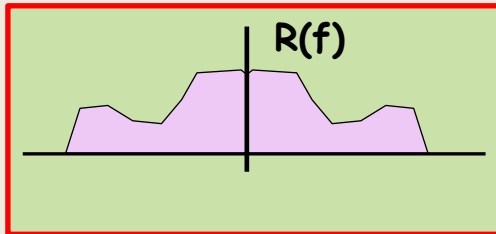
$$\begin{aligned} \text{Capacity}(|H(f)|^2, R_N(f), R(f)) &= \int_0^\infty \log_2 \left( 1 + \frac{R(f)|H(f)|^2}{R_N(f)} \right) df \\ &= \frac{1}{2} \int_{-\infty}^\infty \log_2 \left( 1 + \frac{R(f)|H(f)|^2}{R_N(f)} \right) df \end{aligned}$$





# Lecture 4: Capacity

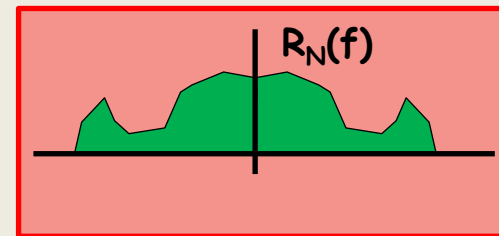
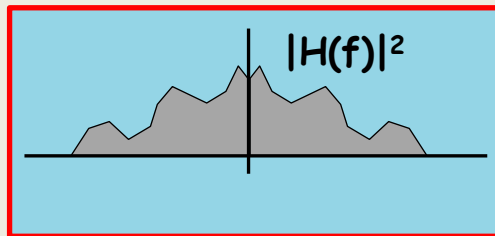
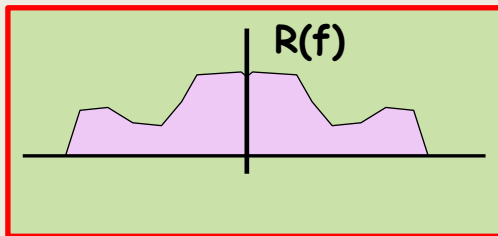
$$C = \max_{R(f): \int R(f) df = P} \frac{1}{2} \int_{-\infty}^{\infty} \log_2 \left( 1 + \frac{R(f)|H(f)|^2}{R_N(f)} \right) df$$



# Lecture 4: Capacity

How to solve the below problem? WATERFILLING

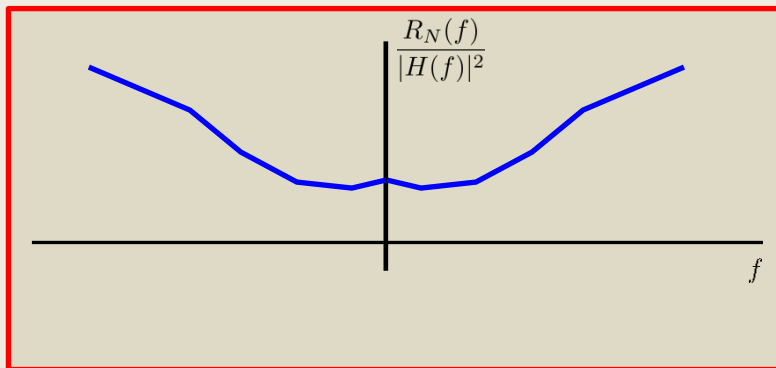
$$C = \max_{R(f): \int R(f) df = P} \frac{1}{2} \int_{-\infty}^{\infty} \log_2 \left( 1 + \frac{R(f)|H(f)|^2}{R_N(f)} \right) df$$



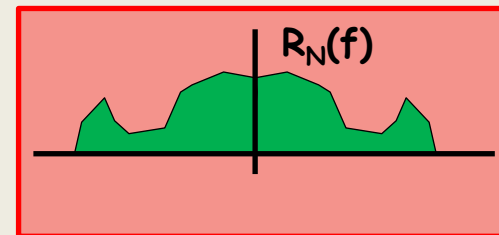
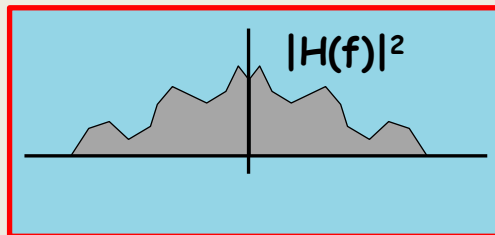
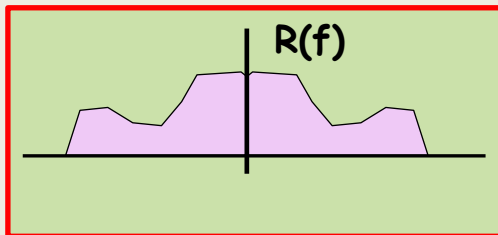
# Lecture 4: Capacity

How to solve the below problem? WATERFILLING

Step 1. Find and plot  $\frac{R_N(f)}{|H(f)|^2}$



$$C = \max_{R(f): \int R(f) df = P} \frac{1}{2} \int_{-\infty}^{\infty} \log_2 \left( 1 + \frac{R(f)|H(f)|^2}{R_N(f)} \right) df$$

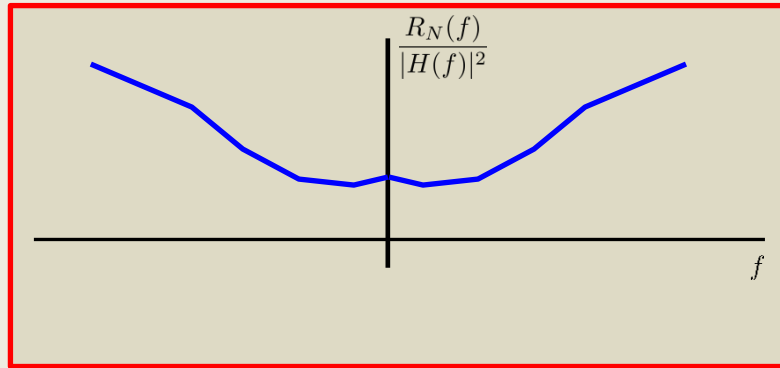


# Lecture 4: Capacity

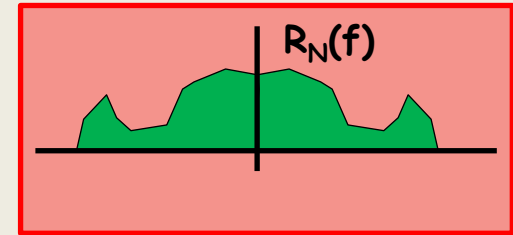
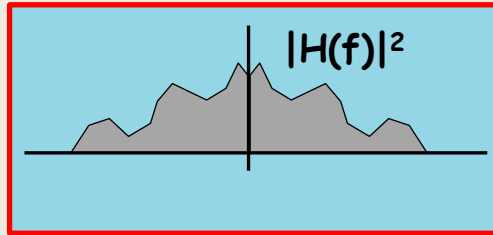
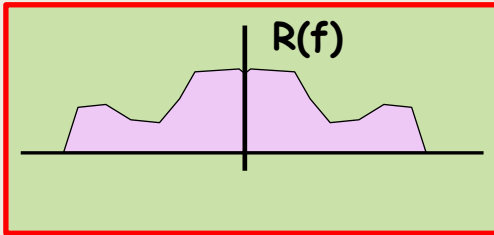
How to solve the below problem? WATERFILLING

Step 1. Find and plot  $\frac{R_N(f)}{|H(f)|^2}$

Step 2. Fill a bucket with  $P$  units of water



$$C = \max_{R(f): \int R(f) df = P} \frac{1}{2} \int_{-\infty}^{\infty} \log_2 \left( 1 + \frac{R(f)|H(f)|^2}{R_N(f)} \right) df$$



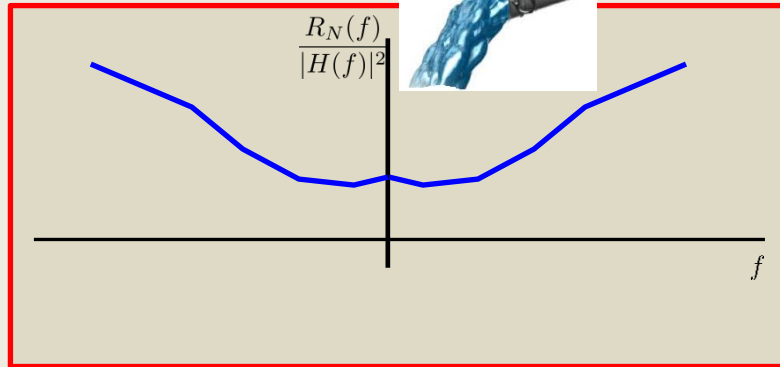
# Lecture 4: Capacity

How to solve the below problem? WATERFILLING

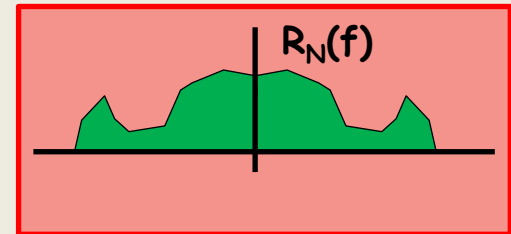
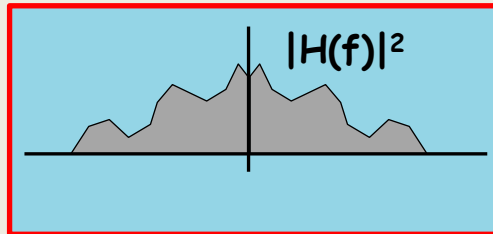
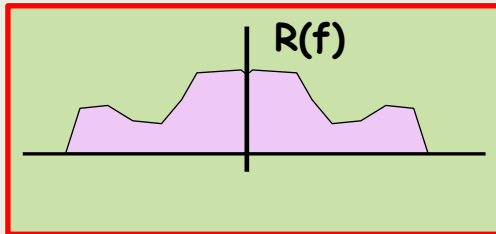
Step 1. Find and plot  $\frac{R_N(f)}{|H(f)|^2}$

Step 2. Fill a bucket with  $P$  units of water

Step 3. Pour it in the shape



$$C = \max_{R(f): \int R(f) df = P} \frac{1}{2} \int_{-\infty}^{\infty} \log_2 \left( 1 + \frac{R(f)|H(f)|^2}{R_N(f)} \right) df$$



# Lecture 4: Capacity

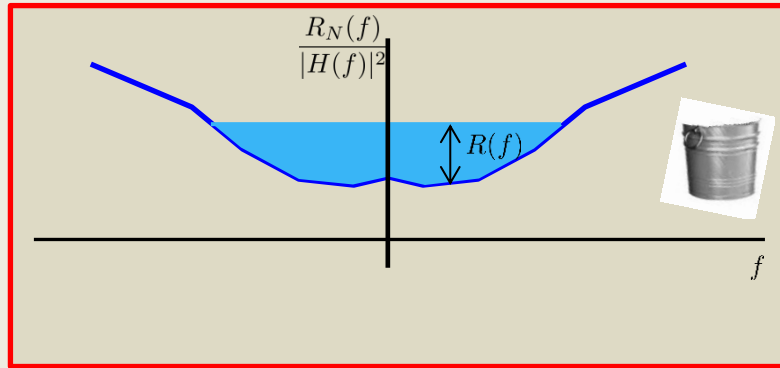
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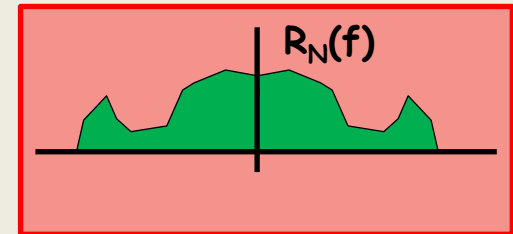
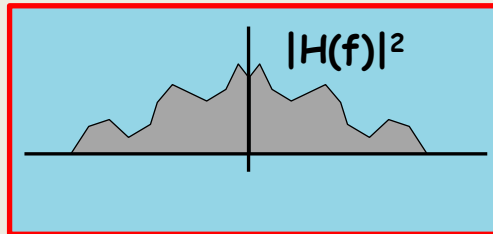
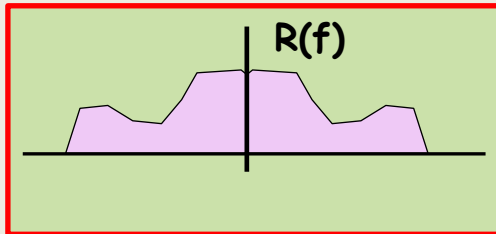


Step 3. Pour it in the shape



Step 4.  
 $R(f)$  is the  
water-level

$$C = \max_{R(f): \int R(f) df = P} \frac{1}{2} \int_{-\infty}^{\infty} \log_2 \left( 1 + \frac{R(f)|H(f)|^2}{R_N(f)} \right) df$$



# Lecture 4: Capacity

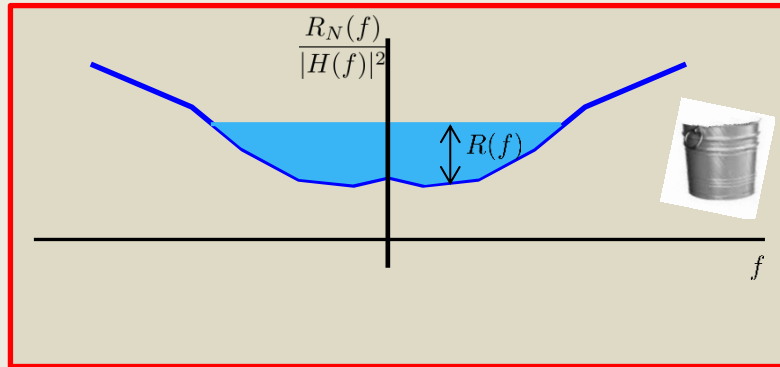
How to solve the below problem? WATERFILLING

Step 1. Find and plot  $\frac{R_N(f)}{|H(f)|^2}$

Step 2. Fill a bucket with  $P$  units of water

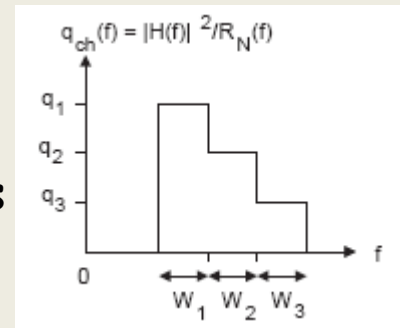


Step 3. Pour it in the shape



Step 4.  
 $R(f)$  is the  
water-level

On Exam,  $|H(f)|^2$   
would be "nice", such as



# Lecture 4: Capacity

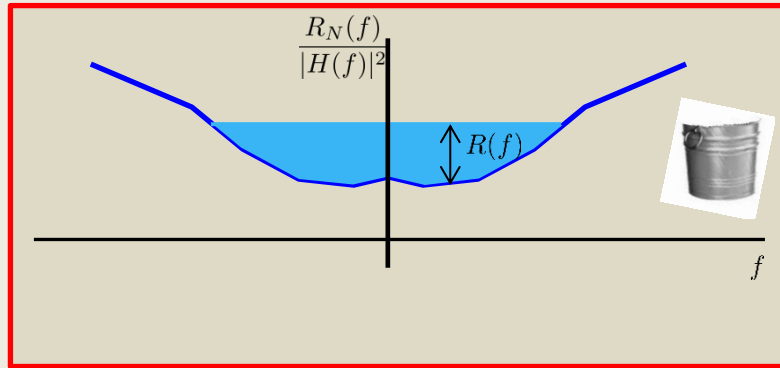
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Step 1. Find and plot  $\frac{R_N(f)}{|H(f)|^2}$

Step 2. Fill a bucket with P units of water



Step 3. Pour it in the shape



Step 4.  
 $R(f)$  is the  
water-level

Observations:

1. Good channels get more power than bad
2. At very high SNRs, all channels get, roughly, the same power