



LUND  
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# Information Theory

Lecture 11

Gaussian channel

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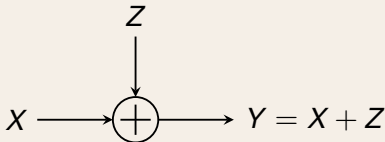
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# Gaussian channel

## Definition

A **Gaussian channel** is a time-discrete channel with input  $X$  and output  $Y = X + Z$ , where  $Z$  models the noise and is  $\mathcal{N}(0, \sqrt{N})$ .



The communication is limited by a power constraint on the transmitter side,

$$E[X^2] \leq P$$

# Capacity

## Definition

The **information capacity** for a continuous channel is

$$C = \max_{\substack{E[X^2]=P \\ f(x)}} I(X; Y)$$

## Theorem

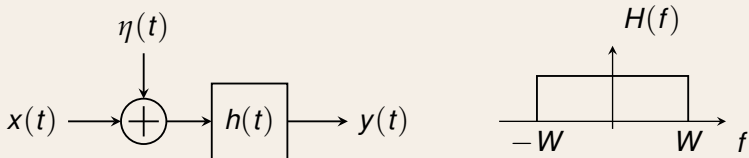
*The mutual information for a Gaussian channel is maximized for  $X \sim \mathcal{N}(0, \sqrt{P})$ , as*

$$C = \max_{\substack{E[X^2]=P \\ f(x)}} I(X; Y) = \max_{\substack{E[X^2]=P \\ f(x)}} H(Y) - H(Z) = \frac{1}{2} \log \left( 1 + \frac{P}{N} \right)$$

# Band limited signal

## Definition (AWGN)

A band limited Gaussian channel consists of a band limited input  $x(t)$ , where  $f_{\max} = W$ , additive white Gaussian noise with  $R(f) = N_0/2$ , and an ideal low-pass filter at the receiver.



The noise at the receiver, after filter and sampling, has the distribution  $Z \sim \mathcal{N}(0, \sqrt{N_0/2})$ .

# Sampling theorem

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## Theorem

Let  $x(t)$  be a band limited signal,  $f_{\max} \leq W$ . If the signal is sampled with  $F_s = 2W$  samples per second to form the sequence  $x[k] = x(\frac{k}{2W})$ , it can be reconstructed with

$$x(t) = \sum_{k=-\infty}^{\infty} x\left(\frac{k}{2W}\right) \operatorname{sinc}\left(t - \frac{k}{2W}\right)$$

where

$$\operatorname{sinc}(t) = \frac{\sin(2\pi Wt)}{2\pi Wt}$$

The sinc-function is 1 for  $t = 0$  and 0 for  $t = k/2W$ ,  $k \neq 0$ .

# Capacity for band limited signals

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## Theorem

Let  $x(t)$  be a band limited signal,  $f_{\max} \leq W$ , and  $z(t)$  noise with power spectra  $R(f) = N_0/2, |f| \leq W$ . The channel

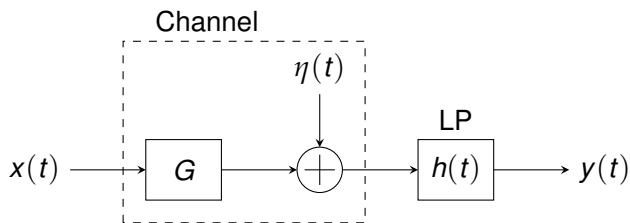
$$y(t) = x(t) + z(t)$$

has the capacity

$$C = W \log\left(1 + \frac{P}{N_0 W}\right) \quad [\text{b/s}]$$

# Attenuated channel

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where  $G$  is the attenuation on the channel ( $|G| < 1$ ) which reduces the received power. The capacity is:

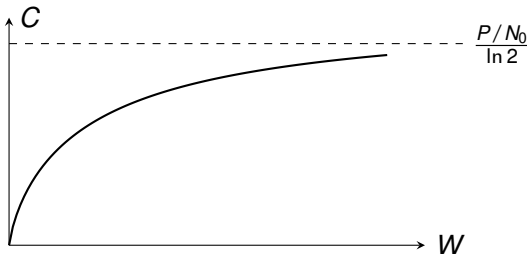
$$C = W \log \left( 1 + \frac{P|G|^2}{N_0 W} \right)$$

# Fundamental limit

## Limit of capacity

As the bandwidth  $W$  grows the capacity approaches a limit,

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





# Fundamental limit

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## Theorem (Shannon limit)

*For reliable communication over white Gaussian noise it is required that*

$$\frac{E_b}{N_0} > \ln 2 = 0.69 = -1.59 \text{ dB}$$

I.e. reliable communication is possible *if and only if*  $\frac{E_b}{N_0} > -1.6\text{dB}$ .  
But what does it require?



# Fundamental limit

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## Theorem

*The maximum achievable bandwidth efficiency is limited by*

$$\frac{C}{W} < \log\left(1 + \frac{C}{W} \cdot \frac{E_b}{N_0}\right)$$

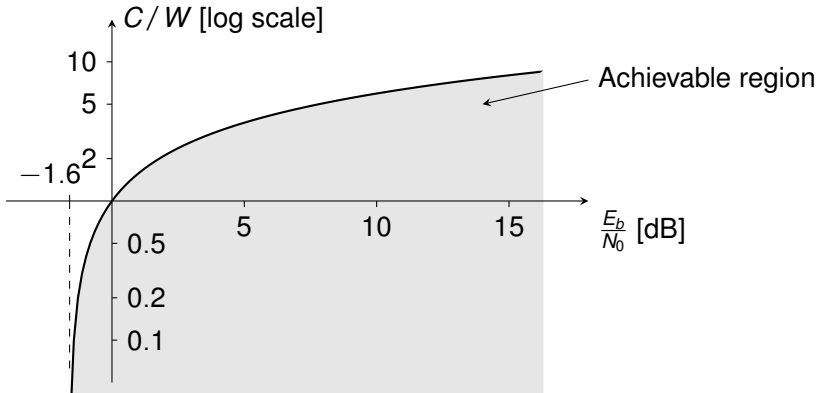
## Corollary

*For reliable communication over white Gaussian noise it is required that*

$$\frac{E_b}{N_0} > \frac{2^{C/W} - 1}{C/W}$$

# Fundamental limit

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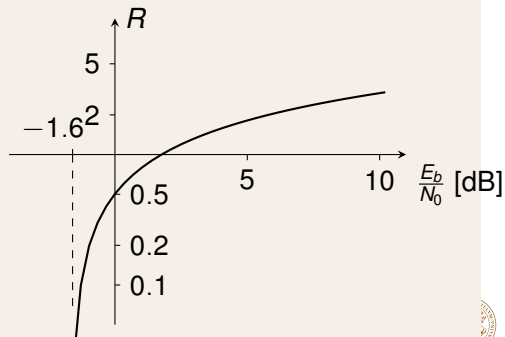


# Fundamental limit

## Theorem

*For reliable communication over white Gaussian noise and a code rate of  $R$  it is required that*

$$\frac{E_b}{N_0} > \frac{2^{2R} - 1}{2R}$$



# Fundamental limit

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## Limit of capacity

To reach the reliable communication as  $\frac{E_b}{N_0} \rightarrow -1.6$  dB the bandwidth efficiency and the code rate approach zero. That indicates a system where

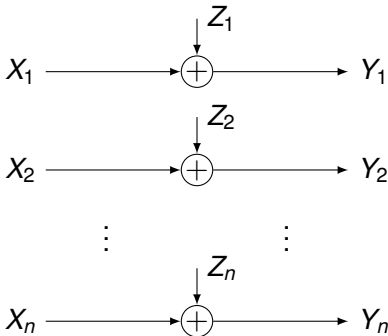
- Codeword length grows to infinity (for finite information word)
- Computational complexity grows to infinity



# Parallel Gaussian channels

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Consider  $n$  parallel, independent, Gaussian channels with a total power constraint of  $\sum_i P_i \leq P$ . The noise on channel  $i$  is  $Z_i \sim \mathcal{N}(0, \sqrt{N_i})$ .



# Parallel Gaussian channels

## Water filling

### Theorem

Given  $k$  independent parallel Gaussian channels with noise variance  $N_i$ ,  $i = 1, \dots, k$ , and a restricted total transmitted power,  $\sum_i P_i = P$ .

The capacity is given by

$$C = \frac{1}{2} \sum_{i=1}^n \log\left(1 + \frac{P_i}{N_i}\right)$$

where

$$P_i = (B - N_i)^+, \quad (x)^+ = \begin{cases} x, & x \geq 0 \\ 0, & x < 0 \end{cases}$$

and  $B$  is such that  $\sum_i P_i = P$ .