

## Summary

- Frequency-domain representation of discrete signals and systems
  - Response of an LTI system to a complex exponential
  - Fourier representation of a discrete-time sequence
- A Review of the discrete-time Fourier Transform (DTFT)
  - Symmetry properties of the Fourier Transform
  - Theorems regarding the Fourier Transform
  - Table of Fourier pairs
- The DTFT of the auto-correlation and of the cross-correlation
  - the DTFT of the auto-correlation
  - the DTFT of the cross-correlation
  - examples



• Question: what is the output of an LTI system when the input is a complex exponential ?  $x[n] = e^{j\omega n}$ ,  $-\infty < n < +\infty$ 

$$y[n] = \sum_{k=-\infty}^{+\infty} x[n]h[n-k] = \sum_{k=-\infty}^{+\infty} h[k]x[n-k] = \sum_{k=-\infty}^{+\infty} h[k]e^{j\omega(n-k)} = \sum_{k=-\infty}^{+\infty} h[k]e^{-j\omega k}e^{j\omega n} = H(e^{j\omega})e^{j\omega n}$$

- Answer: it's the complex exponential possibly modified in magnitude and phase according to the <u>frequency response</u> of the LTI system.
- **Note**: this result reveals that  $e^{j\omega n}$  is an eigen function of the LTI system and that  $H(e^{j\omega})$  is the eigen value of the system at the angular frequency  $\omega$  radians.
- Definition of the frequency response of an LTI system (obtained by computing the Fourier transform of its impulse response)

$$H\left(e^{j\omega}\right)^{\Delta} = \sum_{n=-\infty}^{+\infty} h[n]e^{-j\omega n} = \left|H\left(e^{j\omega}\right)\right|e^{j\angle H\left(e^{j\omega}\right)}$$

- $-~|H(e^{j\omega})|~~\rightarrow~$  absolute value of the frequency response of the system
- $\ \angle H(e^{j\omega}) \ \rightarrow \ phase of the frequency response of the system$



## Frequency-domain representation of discrete signals & systems

Example: what is the response of an LTI system, with h[n] real, to the input x[n]=Acos(ω<sub>0</sub>n+φ) ?

- **Answer**: x[n] may be expressed in a convenient way:  $x[n] = \frac{A}{2} \left[ e^{j(\omega_0 n + \phi)} + e^{-j(\omega_0 n + \phi)} \right]$ and then:

$$y[n] = \frac{A}{2} \Big[ H(e^{j\omega_0}) e^{j(\omega_0 n + \phi)} + H(e^{-j\omega_0}) e^{-j(\omega_0 n + \phi)} \Big] = A \Big| H(e^{j\omega_0}) \cos[\omega_0 n + \phi + \angle H(e^{j\omega_0})] \Big]$$

– Important property of  $H(e^{j\omega})$ 

given the <u>periodicity</u> of the discrete complex exponential,  $e^{j\omega n}$ , the frequency response  $H(e^{j\omega})$  is periodic with period  $2\pi$ , so that in order to characterize it completely, it is sufficient to represent the magnitude and phase considering a frequency span of  $2\pi$  radians, e.g., between  $-\pi$  and  $+\pi$  or 0 and  $2\pi$ .

Example: what is the frequency response of a moving-average filter of length 5 ?

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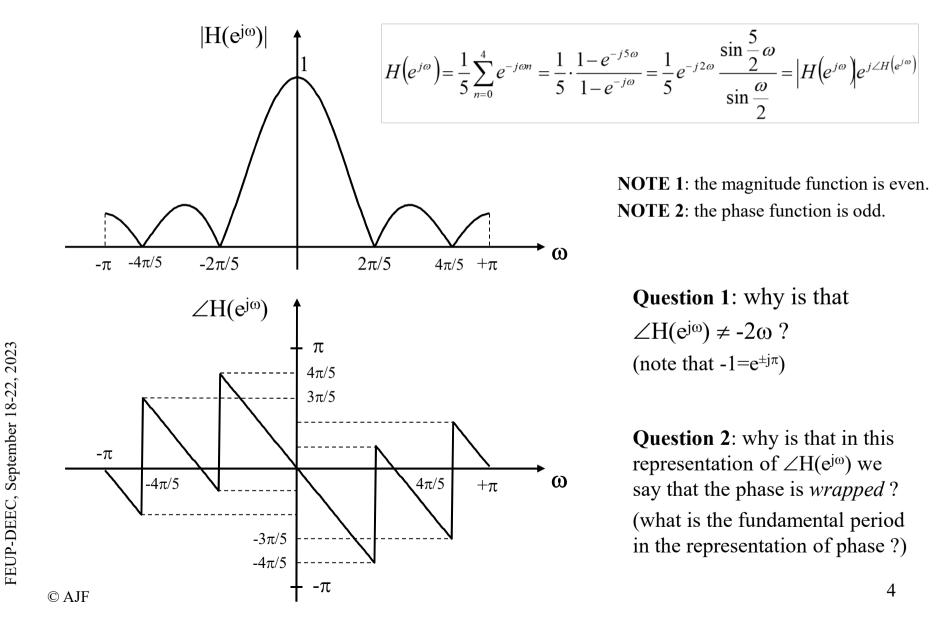
$$h[n] = \begin{cases} 1/5 & 0 \le n \le 4 \\ 0 & outros \end{cases} \qquad \cdots \qquad \underbrace{1/5}_{-3 - 2 - 1 \ 0 \ 1 \ 2 \ 3 \ 4 \ n}$$



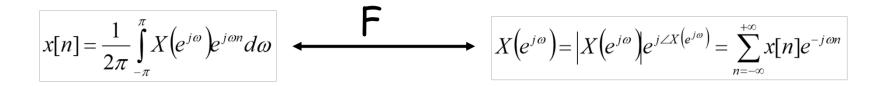
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Frequency-domain representation of discrete signals & systems

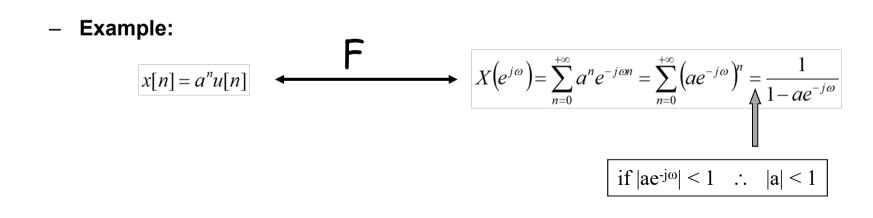
- **Answer**: using the definition of the time-discrete Fourier transform:





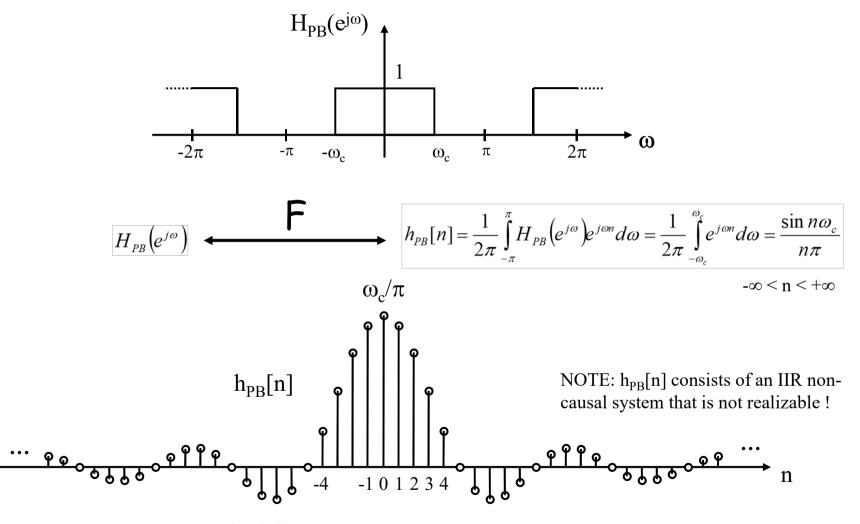


- the Fourier transform of a discrete-time signal x[n] is periodic with period  $2\pi$  and exists if x[n] is absolutely summable
- the inverse Fourier transform allows to synthesize x[n] using a period of its representation in the frequency domain





• Example: what is the impulse response of an ideal low-pass filter ?

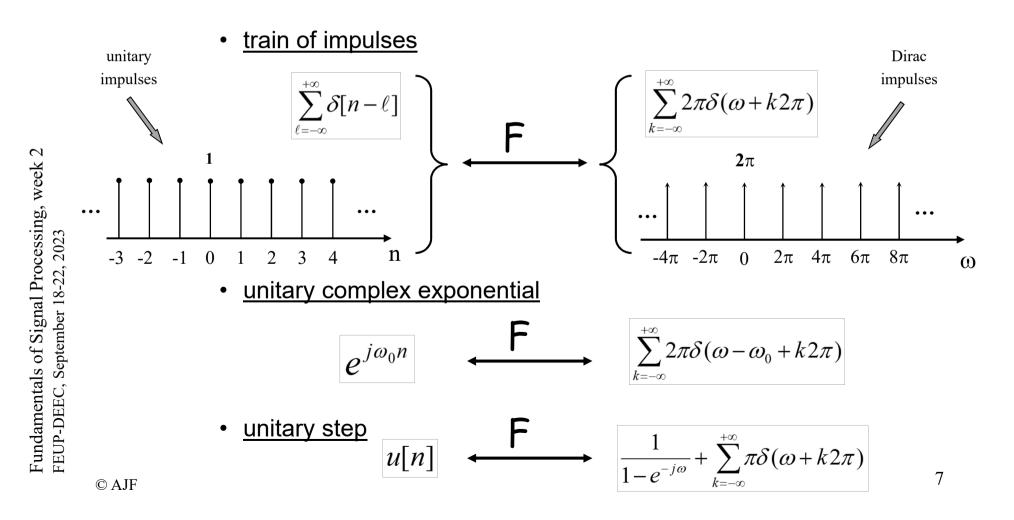


NOTE+: the response  $h_{PB}[n]$  is not absolutely summable, but its square is summable, which highlights the fact that a filter resulting fom  $h_{PB}[n]$  by limiting its length, is the best approximation, in the mean-square sense, to  $H_{PB}(e^{j\omega})$  (*i.e.* to the ideal filter).



- special cases

these are special cases because they are neither absolutely summable nor square-summable, they arise from the theory of generalized functions but they are very important in the analysis of signals and discrete-time systems:





- given x[n], we may express  $x[n]=x_e[n]+x_o[n]$  where:

$$x_e[n] = \frac{1}{2} (x[n] + x^*[-n]) = x_e^*[-n]$$

X<sub>e</sub>[n] is the <u>conjugate symmetric sequence</u> of x[n]; in case x[n] is real,
X<sub>e</sub>[n] is also known as the *even* component of x[n] since X<sub>e</sub>[n]= x<sub>e</sub>[-n]

$$x_o[n] = \frac{1}{2} \left( x[n] - x^*[-n] \right) = -x_o^*[-n]$$

x<sub>o</sub>[n] is the <u>conjugate anti-symmetric sequence of x[n]</u>; in case x[n] is real,
x<sub>o</sub>[n] is also known as the *odd* component of x[n] since x<sub>o</sub>[n]= -x<sub>o</sub>[-n]

- similarly, 
$$X(e^{j\omega}) = X_e(e^{j\omega}) + X_o(e^{j\omega})$$

$$X_e(e^{j\omega}) = \frac{1}{2} \left[ X(e^{j\omega}) + X^*(e^{-j\omega}) \right] = X_e^*(e^{-j\omega})$$

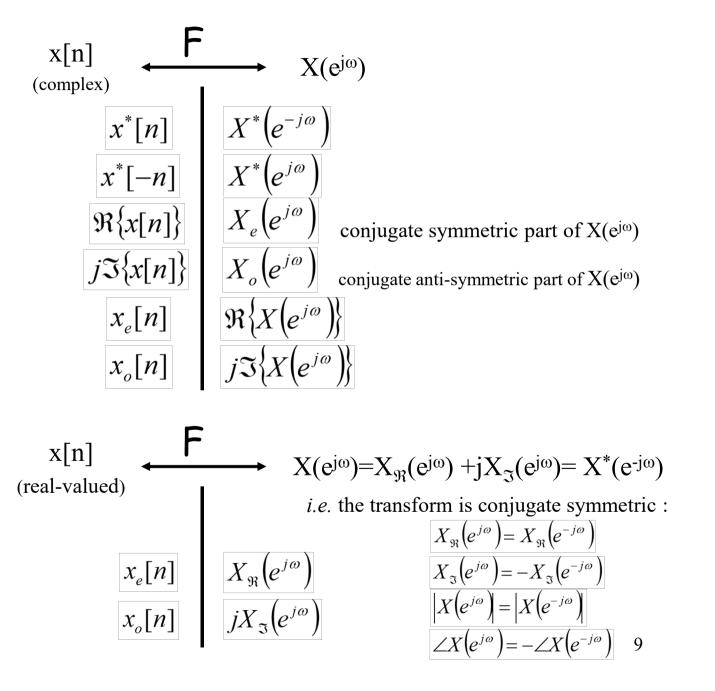
 X<sub>e</sub>(e<sup>jω</sup>) is the <u>conjugate symmetric function</u> of X(e<sup>jω</sup>), X<sub>e</sub>(e<sup>jω</sup>) is also said the *even* component of X(e<sup>jω</sup>) when X(e<sup>jω</sup>) is real-valued

$$X_{o}\left(e^{j\omega}\right) = \frac{1}{2}\left[X\left(e^{j\omega}\right) - X^{*}\left(e^{-j\omega}\right)\right] = -X_{o}^{*}\left(e^{-j\omega}\right)$$

•  $X_o(e^{j\omega})$  is the <u>conjugate anti-symmetric function</u> of  $X(e^{j\omega})$ ,  $X_o(e^{j\omega})$  is also said the *odd* component of  $X(e^{j\omega})$  when  $X(e^{j\omega})$  is real-valued 8

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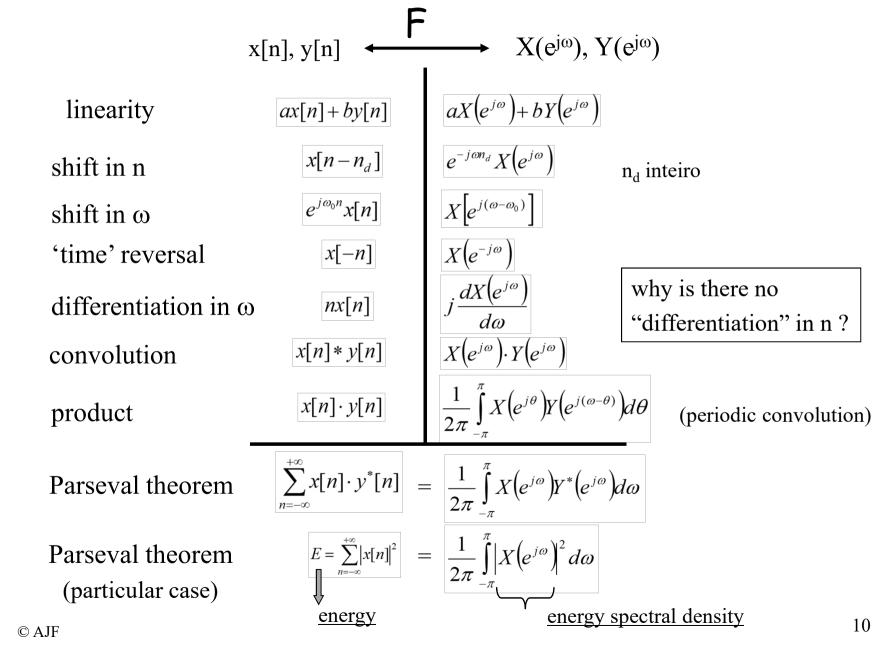


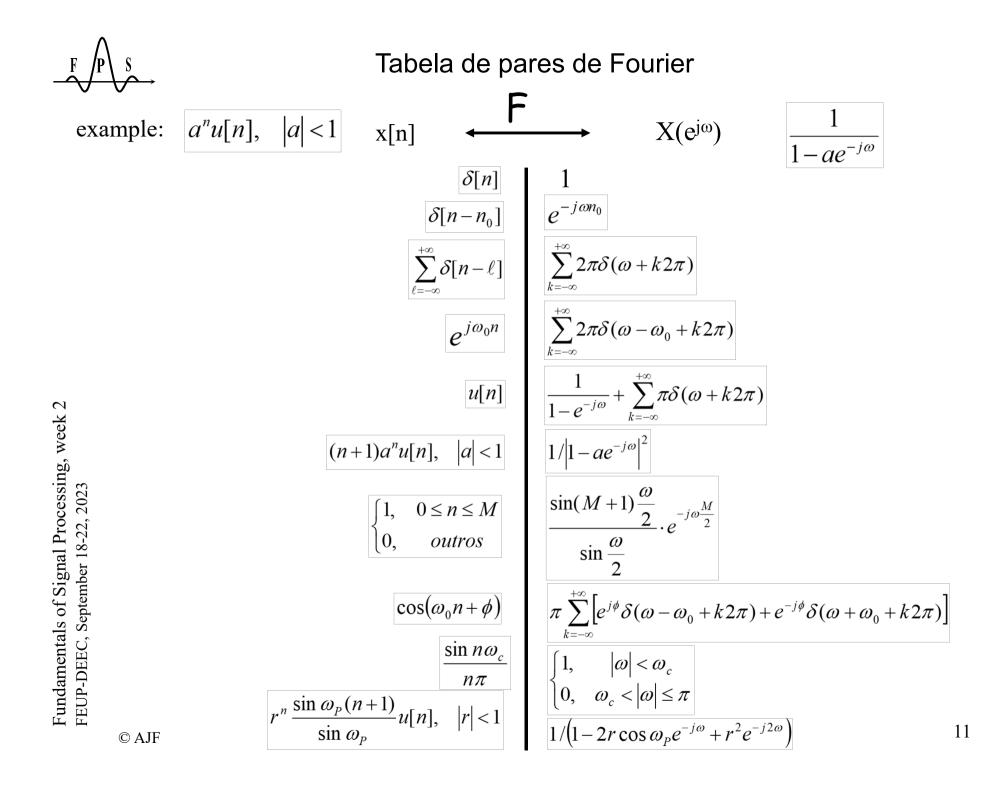




## Review of the main Fourier transform theorems

(relate operations involving discrete sequences and the corresponding operations in the Fourier domain)







Question: what is a practical way to find the inverse Fourier transform ?

• Example:  $X(e^{j\omega}) = \frac{1}{(1 - ae^{-j\omega})(1 - be^{-j\omega})}$ , causal  $\leftarrow$ x[n] = ?if M<N and poles are first-order, then:  $X(e^{j\omega}) = \frac{\prod_{\ell=1}^{M} (1 - c_{\ell} e^{-j\omega})}{\prod_{k=1}^{N} (1 - d_{k} e^{-j\omega})} = \sum_{k=1}^{N} \frac{A_{k}}{1 - d_{k} e^{-j\omega}}$ with :  $A_k = (1 - d_k e^{-j\omega}) X(e^{j\omega}) \Big|_{e^{j\omega} = d_k}$ and thus:  $\frac{1}{(1-ae^{-j\omega})(1-be^{-j\omega})} = \frac{a/(a-b)}{1-ae^{-j\omega}} + \frac{b/(b-a)}{1-be^{-j\omega}}$ which leads to:  $x(n) = \frac{a}{a-b}a^{n}u[n] + \frac{b}{b-a}b^{n}u[n]$ Not to forget !



• the DTFT of the auto-correlation

the auto-correlation is defined as (in this discussion, we admit energy signals)

$$r_{x}[\ell] = x[\ell] * x^{*}[-\ell] = \sum_{k=-\infty}^{+\infty} x[k] x^{*}[k-\ell]$$

considering the DTFT properties

$$\begin{array}{cccc} x[\ell] & \stackrel{\mathcal{F}}{\longleftrightarrow} & X(e^{j\omega}) \\ x^*[\ell] & \stackrel{\mathcal{F}}{\longleftrightarrow} & X^*(e^{-j\omega}) \\ x[-\ell] & \stackrel{\mathcal{F}}{\longleftrightarrow} & X(e^{-j\omega}) \\ x^*[-\ell] & \stackrel{\mathcal{F}}{\longleftrightarrow} & X^*(e^{j\omega}) \end{array}$$

then

$$r_{x}[\ell] = x[\ell] * x^{*}[-\ell] \quad \stackrel{\mathcal{F}}{\longleftrightarrow} \quad R_{x}(e^{j\omega}) = X(e^{j\omega}) \cdot X^{*}(e^{j\omega}) = |X(e^{j\omega})|^{2}$$

Where  $R_x(e^{j\omega}) = |X(e^{j\omega})|^2$  is called the spectral density of energy



- the DTFT of the auto-correlation (cont.)
  - the Wiener-Khinchine Theorem: the auto-correlation and the spectral density of energy form a Fourier pair

$$r_{x}[\ell] \quad \stackrel{\mathcal{F}}{\longleftrightarrow} \quad R_{x}(e^{j\omega}) = |X(e^{j\omega})|^{2}$$

thus,

$$r_{x}[\ell] = \frac{1}{2\pi} \int_{-\pi}^{\pi} R(e^{j\omega}) e^{j\omega\ell} d\omega$$

and, in particular, the energy of the signal can be found using

$$E = r_{x}[0] = \sum_{k=-\infty}^{+\infty} |x[k]|^{2} = \frac{1}{2\pi} \int_{-\pi}^{\pi} R(e^{j\omega}) d\omega = \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(e^{j\omega})|^{2} d\omega$$

which reflects the Parseval Theorem

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the DTFT of the cross-correlation

the cross-correlation is defined as (we admit energy signals)

$$r_{xy}[\ell] = x[\ell] * y^*[-\ell] = \sum_{k=-\infty}^{+\infty} x[k] y^*[k-\ell]$$

considering the DTFT properties

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then

$$r_{xy}[\ell] = x[\ell] * y^*[-\ell] \quad \stackrel{\mathcal{F}}{\longleftrightarrow} \quad R_{xy}(e^{j\omega}) = X(e^{j\omega}) \cdot Y^*(e^{j\omega})$$



• examples

let us admit two discrete-time signals, x[n] and y[n]



it can be easily concluded that

$$\begin{aligned} x[\ell] &= 3\delta[\ell] + 2\delta[\ell-1] + \delta[\ell-2] & \stackrel{\mathcal{F}}{\longleftrightarrow} X(e^{j\omega}) = 3 + 2e^{-j\omega} + e^{-j2\omega} \\ y[\ell] &= \delta[\ell] + 2\delta[\ell-1] + 3\delta[\ell-2] & \stackrel{\mathcal{F}}{\longleftrightarrow} Y(e^{j\omega}) = 1 + 2e^{-j\omega} + 3e^{-j2\omega} \end{aligned}$$

$$R_{x}(e^{j\omega}) = 3e^{j2\omega} + 8e^{j\omega} + 14 + 8e^{-j\omega} + 3e^{-j2\omega} = R_{y}(e^{j\omega}), \text{ (why ?)}$$

$$R_{xy}(e^{j\omega}) = 9e^{j2\omega} + 12e^{j\omega} + 10 + 4e^{-j\omega} + e^{-j2\omega}$$