

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

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

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Frequency-domain representation of discrete signals & systems

- **Example**: what is the response of an LTI system, with h[n] real, to the input $x[n]=Acos(\omega_0 n+\phi)$?

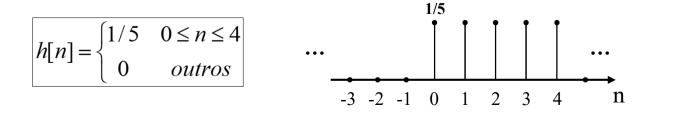
- **Answer**: x[n] may be expressed in a convenient way: $x[n] = \frac{A}{2} \left[e^{j(\omega_0 n + \phi)} + e^{-j(\omega_0 n + \phi)} \right]$ and then:

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

– 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 2π , so that in order to characterize it completely, it is sufficient to represent the magnitude and phase considering a frequency span of 2π radians, e.g., between $-\pi$ and $+\pi$ or 0 and 2π .

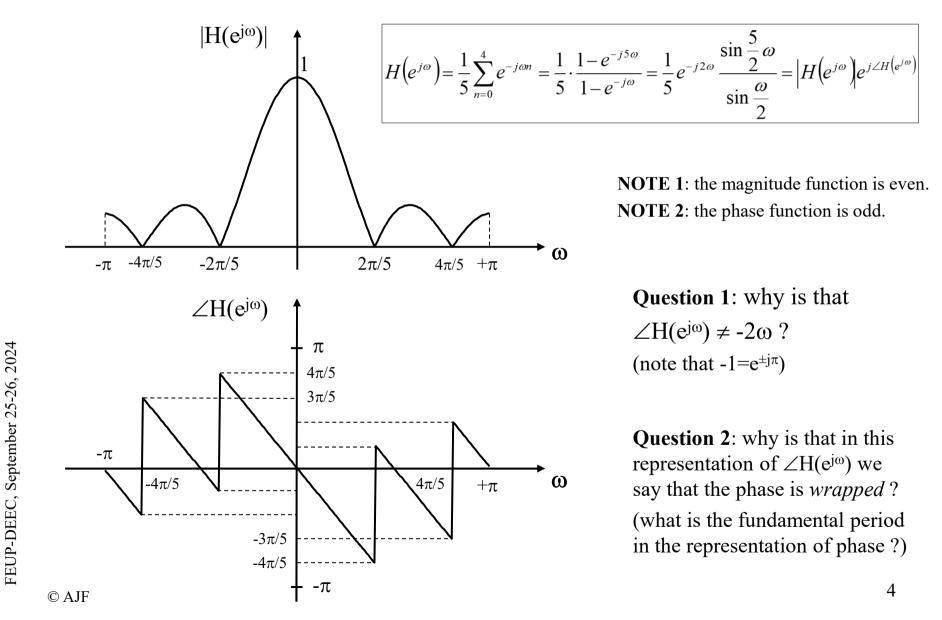
Example: what is the frequency response of a moving-average filter of length 5 ?



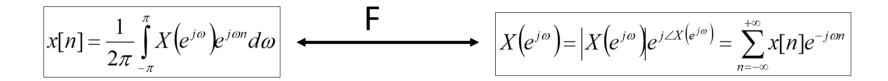


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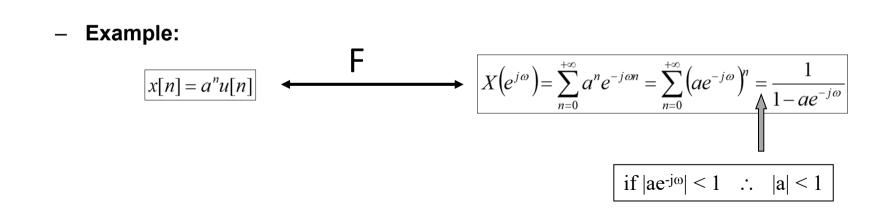
- **Answer**: using the definition of the time-discrete Fourier transform:







- the Fourier transform of a discrete-time signal x[n] is periodic with period 2π 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

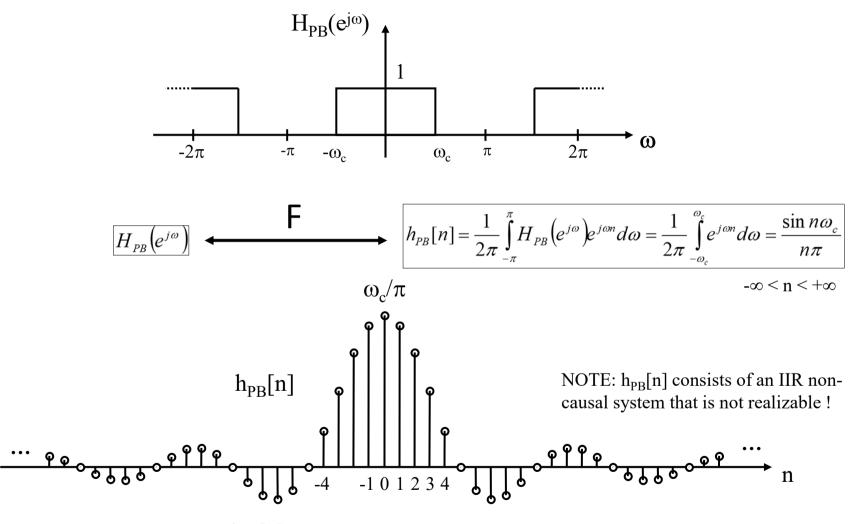


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• Example: what is the impulse response of an ideal low-pass filter ?

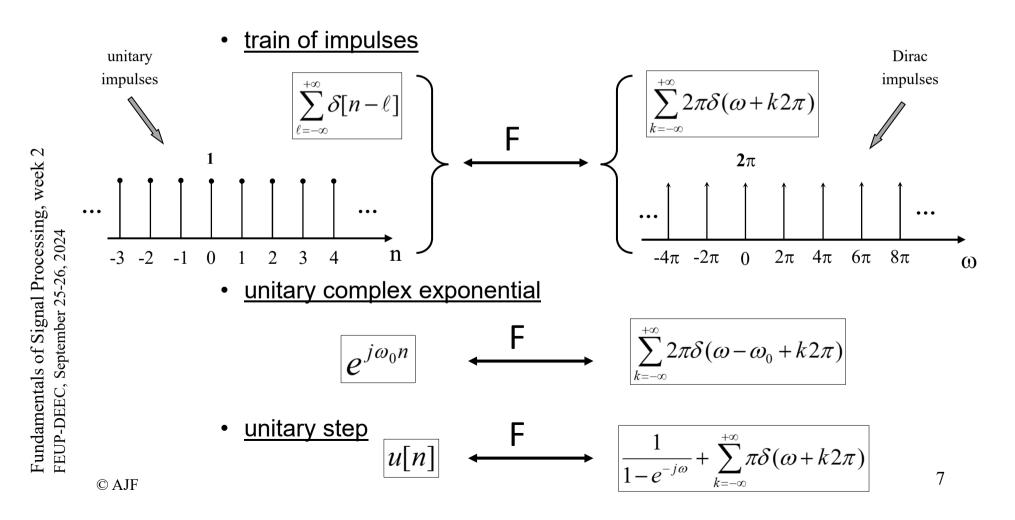


NOTE+: the response $h_{PB}[n]$ is not absolutely summable, but its square is summable, which highlights the fact that a filter resulting fom $h_{PB}[n]$ by limiting its length, is the best approximation, in the mean-square sense, to $H_{PB}(e^{j\omega})$ (*i.e.* to the ideal filter).



- 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:

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

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

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

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

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

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

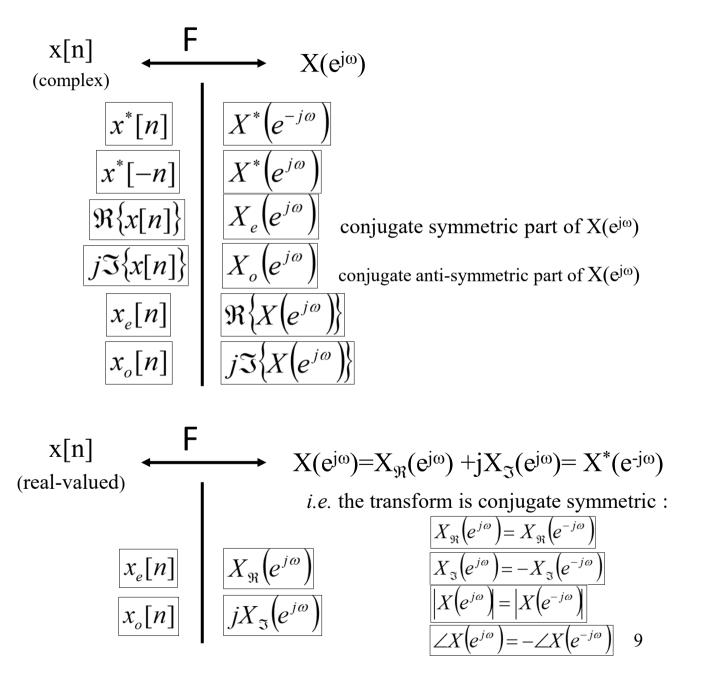
 X_e(e^{jω}) is the <u>conjugate symmetric function</u> of X(e^{jω}), X_e(e^{jω}) is also said the *even* component of X(e^{jω}) when X(e^{jω}) is real-valued

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

• $X_o(e^{j\omega})$ 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



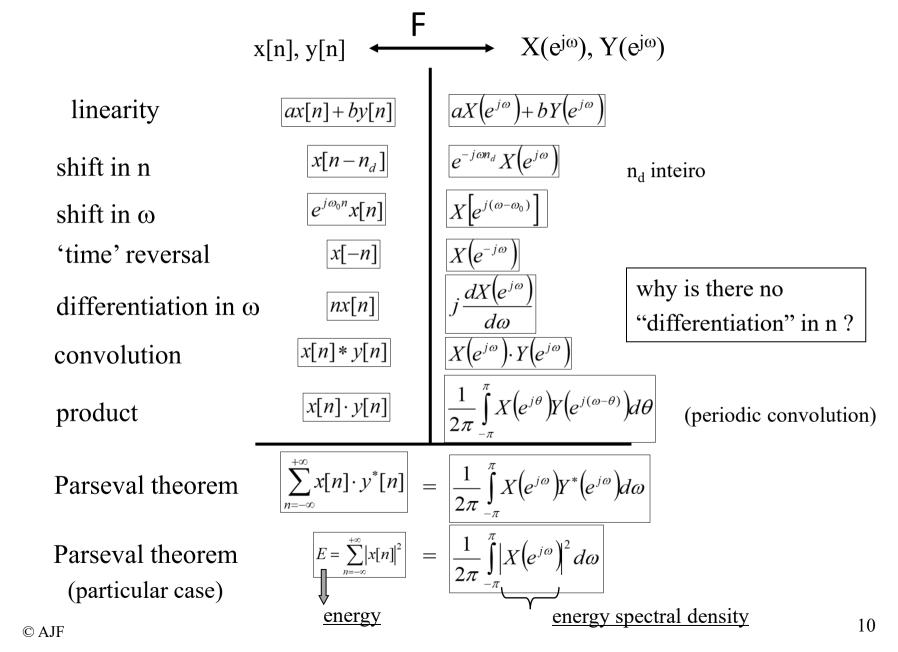
Main symmetry properties of the time-discrete Fourier transform

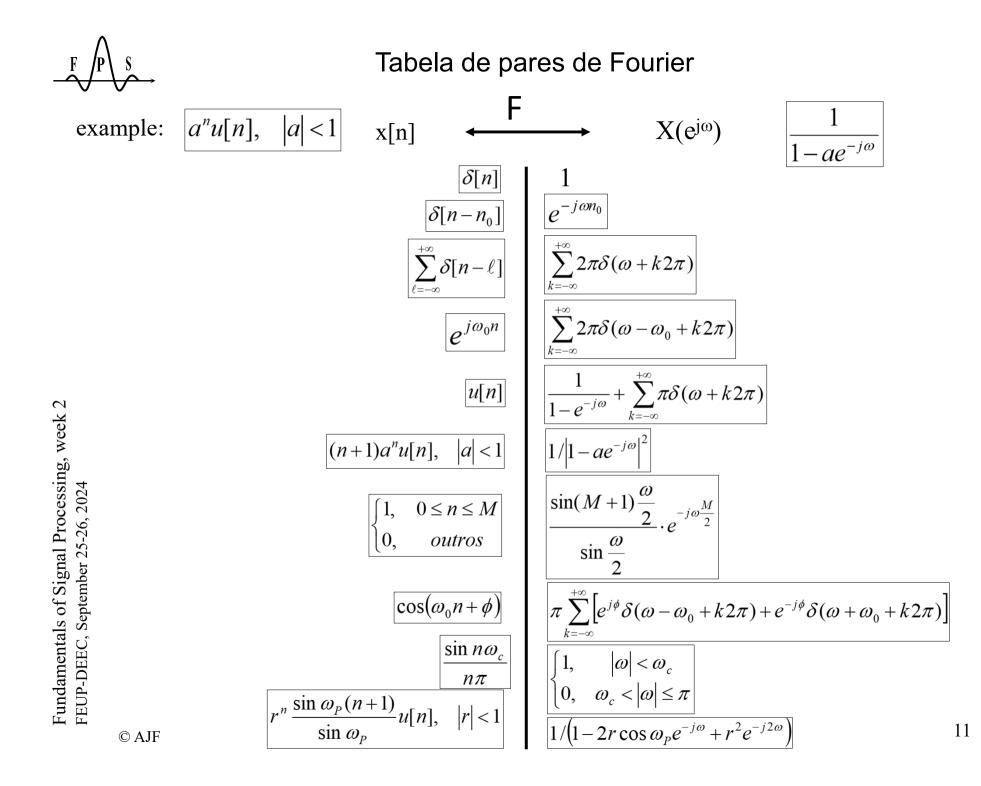




Review of the main Fourier transform theorems

(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 F F x[n] = ? $X(e^{j\omega}) = \frac{\prod_{\ell=1}^{M} (1 - c_{\ell} e^{-j\omega})}{\prod_{k=1}^{N} (1 - d_{k} e^{-j\omega})} = \sum_{k=1}^{N} \frac{A_{k}}{1 - d_{k} e^{-j\omega}}$ if M<N and poles are first-order, then: with : $\left| A_k = (1 - d_k e^{-j\omega}) X(e^{j\omega}) \right|_{e^{j\omega} = d_k}$ 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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$$\begin{array}{cccc} x[\ell] & \stackrel{\mathcal{F}}{\longleftrightarrow} & X(e^{j\omega}) \\ y[\ell] & \stackrel{\mathcal{F}}{\longleftrightarrow} & Y(e^{j\omega}) \\ y^*[\ell] & \stackrel{\mathcal{F}}{\longleftrightarrow} & Y^*(e^{-j\omega}) \\ y[-\ell] & \stackrel{\mathcal{F}}{\longleftrightarrow} & Y(e^{-j\omega}) \\ y^*[-\ell] & \stackrel{\mathcal{F}}{\longleftrightarrow} & Y^*(e^{j\omega}) \end{array}$$

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

then



• examples

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



it can be easily concluded that

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

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

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