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Discrete-time signals in the frequency domain

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• We have observed that each sequence can be expressed in the **time domain** as the **weighted sum of infinite impulse sequences** shifted in time:

$$x(n) = \sum_{m=-\infty}^{+\infty} x(m)\delta(n-m).$$

- In this chapter, we will explore an alternative representation of sequences through the weighted sum of infinite complex exponential sequences of the form $e^{-j\omega n}$, where ω represents the normalized angular frequency.
- This approach allows us to achieve a meaningful representation of sequences in the **frequency domain** and introduces the concept of signal **spectrum**.



- Let us assume that $x_p(t)$ is a complex function $(x_p(t) \in \mathbb{C})$, periodic with period T, continuous in t (with $t \in \mathbb{R}$).
- Then,

$$x_p(t) = \sum_{n=-\infty}^{+\infty} c_n \cdot e^{j\frac{2\pi}{T}nt}$$

where

$$c_n = \frac{1}{T} \int_{-\frac{T}{2}}^{+\frac{T}{2}} f(t) e^{-j\frac{2\pi}{T}nt} \mathrm{d}t$$

• For the Euler's formula

$$e^{j\theta} = \cos\theta + j\,\sin\theta,$$

and $x_p(n)$ is the sum of infinite sine and cosine functions at different frequencies, with each sine/cosine function multiplied by an appropriate weight coefficient.





- The periodic signal is represented in the frequency domain by an infinite number of discrete "lines", corresponding to the coefficients of the Fourier series expansion of the signal.
- These lines are uniformly spaced and separated by $\frac{2\pi}{T}$.



• The frequency domain representation of a continuous-time signal $x_a(t)$ is given by the **Continuous-Time** Fourier Transform (CTFT), defined as

$$X_a(j\Omega) = \int_{-\infty}^{+\infty} x_a(t) e^{-j\Omega t} \mathrm{d}t$$

- The CTFT is also referred to as the Fourier spectrum, or simply spectrum, of the continuous-time signal.
- The continuous-time signal $x_a(t)$ can be reconstructed from its CTFT by means of the **inverse** continuous-time Fourier transform (ICTFT), defined as

$$x_{a}(t) = rac{1}{2\pi} \int_{-\infty}^{+\infty} X_{a}(j\Omega) e^{j\Omega t} \mathrm{d}\Omega.$$



• Note the bijective mapping between the signal $x_a(t)$ and its transform:

$$x_a(t) \stackrel{\mathsf{CTFT}}{\longleftrightarrow} X_a(j\Omega)$$

- Ω is a real variable representing the continuous-time angular frequency, measured in rad/s.
- The inverse transform can be interpreted as the linear combination of infinitesimally small complex exponential signals of the form $\frac{1}{2\pi}e^{j\Omega t}d\Omega$.
- We can also express the transform in polar form:

$$X_a(j\Omega) = |X_a(j\Omega)| \cdot e^{j\theta_a(\Omega)},$$

where $\theta_a(\Omega) = \arg \{X_a(j\Omega)\}.$

- $|X_a(j\Omega)|$ is referred to as the magnitude spectrum, and $\theta_a(\Omega)$ is called the phase spectrum.
- Both $|X_a(j\Omega)|$ and $\theta_a(\Omega)$ are real functions of the angular frequency Ω .



- Note that not all signals admit the CTFT. The integral $\int_{-\infty}^{+\infty} \cdot \mathrm{d}t$ may not converge.
- The CTFT exists if the continuous-time signal $x_a(t)$ satisfies the **Dirichlet conditions**:
 - 1. The signal has a finite number of discontinuities and a finite number of maxima and minima in any finite interval.
 - 2. The signal is absolutely integrable, i.e.,

$$\int_{-\infty}^{+\infty} \left| x_{a}(t)
ight| < +\infty$$

• If these conditions are satisfied, $\int_{-\infty}^{+\infty} x_a(t) e^{-j\Omega t} dt$ converges and $x_a(t) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} X_a(j\Omega) e^{j\Omega t} d\Omega$ apart from the discontinuity points.





• This signal satisfies the Dirichlet conditions: it has a unique discontinuity and

$$\int_{-\infty}^{+\infty} \left| x_{a}(t) \right| = \int_{0}^{+\infty} e^{-\alpha t} \mathrm{d}t = - \left. \frac{e^{-\alpha t}}{\alpha} \right|_{0}^{+\infty} = 0 - \left(-\frac{1}{\alpha} \right) = \frac{1}{\alpha}$$

CTFT Example





Figure 3.2: (a) Magnitude and (b) phase of $X_a(j\Omega) = 1/(0.5/\sec + j\Omega)$.

CTFT of a Dirac delta function

- The **Dirac delta function** $\delta(t)$ is a function of the continuous-time variable *t* with notable properties.
- It is defined as:

$$\delta(t) = \left\{ egin{array}{ccc} 0 & t
eq 0 \ +\infty & t = 0 \end{array}
ight.$$

with

$$\int_{-\infty}^{+\infty} \delta(t) \mathrm{d}t = 1.$$

• It is the limit, as T approaches 0, of the rectangular pulse



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$$\int_{-\infty}^{+\infty} f(t) \,\delta(t) dt = f(0)$$
$$\int_{-\infty}^{+\infty} f(t) \,\delta(t - t_0) dt = f(t_0)$$
$$\mathsf{CTFT}\{\delta(t)\} = \Delta(j\Omega) = \int_{-\infty}^{+\infty} \delta(t) e^{-j\Omega t} dt = 1$$
$$\mathsf{CTFT}\{\delta(t - t_0)\} = \int_{-\infty}^{+\infty} \delta(t - t_0) e^{-j\Omega t} dt = e^{-j\Omega t_0}$$





• Linearity of the CTFT: If $F_a(j\Omega)$ is the CTFT of $f_a(t)$ and $G_a(j\Omega)$ is the CTFT of $g_a(t)$, the CTFT of $x_a(t) = \alpha f_a(t) + \beta g_a(t)$, with α and β constants, is

 $X_{a}(j\Omega) = \alpha F_{a}(j\Omega) + \beta G_{a}(j\Omega).$

• Time-shift property: If $G_a(j\Omega)$ is the CTFT of $g_a(t)$, the CTFT of $x_a(t) = g_a(t - t_0)$, with t_0 constant, is

 $G_a(j\Omega)e^{-j\Omega t_0}.$

• Symmetry property of the CTFT: The CTFT of a real signal $x_a(t) \in \mathbb{R}$ satisfies the property

$$X_a(-j\Omega) = X_a^*(j\Omega)$$

where x^* is the conjugate of x.

• Energy density spectrum: The total energy E_x of a continuous-time signal $x_a(t)$ is given by

$$E_{x}=\int_{-\infty}^{+\infty}\left|x_{a}(t)\right|^{2}\mathrm{d}t.$$



• The total energy E_x of a continuous-time signal $x_a(t)$ is given by

$$E_x = \int_{-\infty}^{+\infty} |x_a(t)|^2 \mathrm{d}t.$$

- An absolutely integrable signal, i.e., a signal for which ∫^{+∞}_{-∞} |x_a(t)|dt < +∞, has finite energy, but there exist signals with finite energy which are not absolutely integrable. Moreover, there are signals with infinite energy (for example, periodic signals).
- If $x_a(t)$ admits $X_a(j\Omega)$, then **Parseval's Theorem** holds:

$$E_{x} = \int_{-\infty}^{+\infty} |x_{a}(t)|^{2} \mathrm{d}t = \frac{1}{2\pi} \int_{-\infty}^{+\infty} |X_{a}(j\Omega)|^{2} \mathrm{d}\Omega$$

- S_{xx}(Ω) = |X_a(jΩ)|² is also called Energy density spectrum of the signal x_a(t), and it provides the energy content of the signal at angular frequency Ω.
- The energy of the signal x_a(t) in a frequency range Ω_a ≤ Ω ≤ Ω_b can be computed by integrating S_{xx}(Ω) over the interval [Ω_a, Ω_b]:

$$\mathcal{E}_{\mathrm{x},r} = rac{1}{2\pi} \int_{\Omega_{m{a}}}^{\Omega_{b}} \mathcal{S}_{\mathrm{xx}}(\Omega) \mathrm{d}\Omega.$$

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- A full-band continuous-time signal $x_a(t)$ has a spectrum occupying the entire frequency range.
- A signal is called **band-limited** if its spectrum occupies only a portion of the frequency range $-\infty < \Omega < +\infty$.
- An ideal band-limited signal has spectrum that is zero outside a certain range $\Omega_a \leq |\Omega| \leq \Omega_b$, with $0 \leq \Omega_a, \Omega_b \leq +\infty$, i.e.,

$$X_{a}(j\Omega) = egin{cases} 0 & 0 \leq |\Omega| < \Omega_{a} \ 0 & \Omega_{b} < |\Omega| < +\infty \end{cases}$$

- It is not possible to generate an ideal band-limited signal, but in most applications, it suffices to ensure that the signal has sufficiently low energy outside the interval [Ω_a, Ω_b] of interest.
- A signal is called **low-pass** if its spectrum occupies the frequency range $0 \le |\Omega| \le \Omega_p$, where Ω_p is called the signal *bandwidth*.
- A signal is called high-pass if its spectrum occupies the frequency range Ω_p ≤ |Ω| ≤ +∞, and the signal bandwidth extends from Ω_p to ∞.
- Eventually, a signal is called **passband** if its spectrum occupies the frequency range $0 < \Omega_L \le |\Omega| \le \Omega_H < +\infty$, where $\Omega_H \Omega_L$ is the signal bandwidth.





- The frequency domain representation of a discrete-time signal is given by the *Discrete-Time Fourier Transform* (DTFT).
- This transform expresses a sequence as a weighted combination of complex exponential sequences of the form e^{jωn}, where ω is the normalized angular frequency, and ω ∈ ℝ.
- If it exists, the DTFT of a sequence is unique and the original sequence can be recovered from its transform with an inverse transform operation.
- The DTFT $X(e^{j\omega})$ of a sequence x(n) is defined by:

$$X(e^{j\omega}) = \sum_{n=-\infty}^{+\infty} x(n)e^{-j\omega n}.$$



• Let us compute the DTFT of the unit impulse sequence $\delta(n)$,

$$\delta(n) = \begin{cases} 1 & n = 0 \\ 0 & n \neq 0 \end{cases}$$

$$\Delta(e^{j\omega}) = \sum_{n=-\infty}^{+\infty} \delta(n)e^{-j\omega n} = 1 \cdot e^{-j\omega 0} = 1$$

DTFT of a causal exponential sequence



 Let us compute the DTFT of the causal exponential sequence x(n) = αⁿμ(n), where |α| < 1 and μ(n) is the unit step sequence.



$$X(e^{j\omega}) = \sum_{n=-\infty}^{+\infty} \alpha^n \mu(n) e^{-j\omega n} = \sum_{n=0}^{+\infty} \alpha^n e^{-j\omega n} = \sum_{n=0}^{+\infty} \left(\alpha e^{-j\omega}\right)^n = \frac{1}{1 - \alpha e^{-j\omega}}$$

since $|\alpha e^{j\omega}| = |\alpha| < 1.$

• To derive this result, we have used the following identity:

$$\sum_{n=0}^{+\infty}q^n=rac{1}{1-q}\qquad \quad ext{if }|q|<1.$$





• The DTFT is a **periodic function** of ω with a period of 2π :

$$X(e^{j(\omega+2\pi k)}) = \sum_{n=-\infty}^{+\infty} x(n)e^{-j(\omega+2\pi k)n} = \sum_{n=-\infty}^{+\infty} x(n)e^{-j\omega n} \cdot e^{-j2\pi kn} = \sum_{n=-\infty}^{+\infty} x(n)e^{-j\omega n} = X(e^{j\omega}),$$

where we used the fact that $e^{-j2\pi kn} = 1$ for k and n integers.

- Note that $\sum_{n=-\infty}^{+\infty} x(n)e^{-j\omega n}$ is the Fourier series expansion of the periodic function $X(e^{j\omega})$.
- The coefficients of the Fourier series, x(n), can be computed from $X(e^{j\omega})$ using the Fourier integral:

$$x(n) = rac{1}{2\pi} \int_{-\pi}^{+\pi} X(e^{j\omega}) e^{j\omega n} \mathrm{d}\omega,$$

which is the Inverse Discrete-Time Fourier Transform (IDTFT).

• The IDTFT can be interpreted as the linear combination of infinitesimally small complex exponential signals of the form $\frac{1}{2\pi}e^{j\omega n}d\omega$ weighted by the complex function $X(e^{j\omega})$.

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IDTFT proof



• Let us verify that

$$X(n) = rac{1}{2\pi} \int_{-\pi}^{+\pi} X(e^{j\omega}) e^{j\omega n} \mathrm{d}\omega.$$

• By replacing the definition of $X(e^{j\omega})$ with

$$X(e^{j\omega}) = \sum_{l=-\infty}^{+\infty} x(l)e^{-j\omega l}$$

we have

$$x(n) = \frac{1}{2\pi} \int_{-\pi}^{+\pi} \sum_{l=-\infty}^{+\infty} x(l) e^{-j\omega l} e^{j\omega n} d\omega =$$

and by interchanging \int and \sum :

$$=\frac{1}{2\pi}\sum_{l=-\infty}^{+\infty}x(l)\int_{-\pi}^{+\pi}e^{j\omega(n-l)}\mathrm{d}\omega$$

• For computing this integral, we will consider two cases.

IDTFT proof



• If $n \neq l$,

$$\int_{-\pi}^{+\pi} e^{j\omega(n-l)} \mathrm{d}\omega = \int_{-\pi}^{+\pi} \cos\left[\omega(n-l)\right] \mathrm{d}\omega + j \int_{-\pi}^{+\pi} \sin\left[\omega(n-l)\right] \mathrm{d}\omega = \left.\frac{\sin\left[\omega(n-l)\right]}{(n-l)}\right|_{-\pi}^{+\pi} = \frac{2\sin\left[\pi(n-l)\right]}{(n-l)} = 0.$$

• If n = l,

$$\int_{-\pi}^{+\pi} e^{j\omega(n-l)} \mathrm{d}\omega = \int_{-\pi}^{+\pi} e^{j\omega 0} \mathrm{d}\omega = 2\pi.$$

• Thus, in general, it is

$$\int_{-\pi}^{+\pi} e^{j\omega(n-l)} \mathrm{d}\omega = 2\pi\delta(n-l)$$

where $\delta(n)$ is the unit impulse sequence.

• Eventually, we have

$$\frac{1}{2\pi}\sum_{l=-\infty}^{+\infty}x(l)\int_{-\pi}^{+\pi}e^{j\omega(n-l)}\mathrm{d}\omega=\sum_{l=-\infty}^{+\infty}x(l)\delta(n-l)=x(n)$$

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• In general, the DTFT of a sequence is a complex function of the real variable ω :

$$X(e^{j\omega}) = X_{
m re}(e^{j\omega}) + jX_{
m im}(e^{j\omega})$$

with $X_{ ext{re}}(e^{j\omega})$ and $X_{ ext{im}}(e^{j\omega}) \in \mathbb{R}.$

$$\begin{split} X_{\rm re}(e^{j\omega}) &= \frac{1}{2} \left[X(e^{j\omega}) + X^*(e^{j\omega}) \right], \\ X_{\rm im}(e^{j\omega}) &= \frac{1}{2j} \left[X(e^{j\omega}) - X^*(e^{j\omega}) \right]. \end{split}$$

• The DTFT can be expressed in polar form as

$$X(e^{j\omega}) = \left| X(e^{j\omega}) \right| e^{j\theta(\omega)}$$

where $|X(e^{j\omega})|$ is called magnitude spectrum, and $\theta(\omega)$ is called the phase spectrum:

$$heta(\omega) = rctan rac{X_{
m im}(e^{j\omega})}{X_{
m re}(e^{j\omega})}$$

• Note that there is an indetermination of $2\pi k$, with $k \in \mathbb{Z}$, in the knowledge of $\theta(\omega)$.

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• If $x(n) \in \mathbb{R}$:

$$X(e^{-j\omega}) = X^*(e^{j\omega}).$$

In fact:

$$X^*(e^{j\omega}) = \sum_{n=-\infty}^{+\infty} x^*(n) e^{j\omega n} = \sum_{n=-\infty}^{+\infty} x(n) e^{-j(-\omega)n} = X(e^{-j\omega}).$$

- For this relation, if $x(n) \in \mathbb{R}$ we have that
 - $X_{\rm re}(e^{-j\omega}) = X_{\rm re}(e^{j\omega})$ is an even function, $X_{\rm im}(e^{-j\omega}) = -X_{\rm im}(e^{j\omega})$ is an odd function, $\left|X(e^{-j\omega})\right| = \left|X(e^{j\omega})\right|$ is an even function, $\theta(-\omega) = -\theta(\omega)$ is an odd function.
- If x(n) is real and even (x(n) = x(-n)),

$$X(e^{j\omega}) = X_{\rm re}(e^{j\omega})$$
 is an real function.

• If x(n) is real and odd (x(-n) = -x(n)),

 $X(e^{j\omega}) = iX_{im}(e^{j\omega})$ is an imaginary function.

- The series that defines the Fourier transform could or could not converge.
- We say that the DTFT exists if its series converges according to some criteria.
- Sufficient conditions for the existence of the DTFT of a sequence x(n) are the following:
 - x(n) is absolutely summable, i.e.,

$$\sum_{-\infty}^{+\infty} |x(n)| < +\infty$$

In this case, we talk about uniform convergence.

• x(n) is a finite energy signal

$$\sum_{-\infty}^{+\infty}|x(n)|^2<+\infty$$

In this case, we talk about mean square convergence.

• These two conditions are quite restrictive. Note that the first condition is stronger than the second. An absolutely summable signal is always a finite energy signal because:

$$\sum_{-\infty}^{+\infty} |x(n)|^2 \le \left(\sum_{-\infty}^{+\infty} |x(n)|\right)^2$$

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• An absolutely summable signal is the following causal exponential sequence

 $x(n) = \alpha^n \mu(n)$

for $|\alpha| < 1$.

• Indeed,

$$\sum_{n=-\infty}^{+\infty} |\alpha^n \mu(n)| = \sum_{n=0}^{+\infty} |\alpha^n| = \frac{1}{1-|\alpha|} < +\infty.$$

We have already seen that its DTFT is $\frac{1}{1-\alpha e^{-j\omega}}$.



• A signal that is not absolutely summable (but has a DTFT) is the signal with the following ideal low-pass DTFT:



• Let us compute the corresponding signal, which we will encounter frequently.

Example of mean square convergence



• Let us compute the IDTFT:

$$h_{\mathrm{LP}}(n) = rac{1}{2\pi} \int_{-\pi}^{+\pi} H_{\mathrm{LP}}(e^{j\omega}) e^{j\omega n} \mathrm{d}\omega,$$

• For $-\infty < n < +\infty$ and $n \neq 0$:

$$h_{\rm LP}(n) = \frac{1}{2\pi} \int_{-\omega_c}^{+\omega_c} e^{j\omega n} d\omega = \frac{1}{2\pi} \left[\frac{e^{j\omega_c n}}{jn} - \frac{e^{-j\omega_c n}}{jn} \right] = \frac{\sin(\omega_c n)}{\pi n},$$

• For *n* = 0

$$h_{\rm LP}(n) = \frac{1}{2\pi} \int_{-\omega_c}^{+\omega_c} \mathrm{d}\omega = \frac{\omega_c}{\pi}$$

• Thus,

$$h_{\rm LP}(n) = \begin{cases} \frac{\omega_c}{\pi} & n = 0\\ \frac{\sin(\omega_c n)}{\pi n} & n \neq 0 \end{cases}$$

• Since

$$\frac{\sin \omega_c n}{\pi n} = \frac{\omega_c}{\pi} \frac{\sin(\omega_c n)}{\omega_c n},$$

and assuming by convention that $\frac{\sin(\omega_c 0)}{\omega_c 0} = 1$, we can write for all n

$$h_{\rm LP}(n)=\frac{\sin(\omega_c n)}{\pi n}$$

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- The DTFT can be defined also for a certain class of sequences that are not absolutely summable, nor have finite energy, e.g., the unit step sequence, the sinusoidal sequence, or the complex exponential sequence.
- In these cases the expression of the DTFT involves Dirac delta functions.

Example



• The DTFT of the complex exponential sequence $x(n) = e^{j\omega_0 n}$ is

$$X(e^{j\omega}) = \sum_{k=-\infty}^{+\infty} 2\pi \delta(\omega - \omega_0 + 2\pi k)$$

The DTFT $X(e^{j\omega})$ is a periodic function in ω with period 2π and it is called a **periodic pulse train**.



• Let us prove this relation by computing the IDTFT:

$$x(n) = \frac{1}{2\pi} \int_{-\pi}^{+\pi} \sum_{k=-\infty}^{+\infty} 2\pi \delta(\omega - \omega_0 + 2\pi k) e^{j\omega n} \mathrm{d}\omega$$

• If $-\pi < \omega_o \le \pi$, the only non-null function in the interval $[-\pi, +\pi]$ is $\delta(\omega - \omega_0)$, thus

$$x(n) = \frac{1}{2\pi} 2\pi \int_{\text{Digital Signal}}^{+\pi} \delta(\omega - \omega_0) e^{j\omega n} d\omega = e^{j\omega_0 n}$$

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- Let us assume that $G(e^{j\omega})$ is the DTFT of g(n) and that $H(e^{j\omega})$ is the DTFT of h(n).
- Linearity property: If $x(n) = \alpha g(n) + \beta h(n)$, with α and β constants, the DTFT of x(n) is $X(e^{j\omega}) = \alpha G(e^{j\omega}) + \beta H(e^{j\omega})$.
- Time reversal: The DTFT of g(-n) is $G(e^{-j\omega})$.
- Time-shifting: The DTFT of $g(n n_0)$ is $e^{-j\omega n_0} G(e^{j\omega})$.
- Frequency-shifting: The DTFT of $e^{j\omega_0 n}g(n)$ is $G(e^{j(\omega-\omega_0)})$.
- Frequency differentiation: The DTFT of ng(n) is $j \frac{dG(e^{j\omega})}{d\omega}$.
- Modulation: The DTFT of $g(n) \cdot h(n)$ is $\frac{1}{2\pi} \int_{-\pi}^{+\pi} G(e^{j\theta}) H(e^{j(\omega-\theta)}) d\theta$.
- Parseval theorem:

$$\sum_{n=-\infty}^{+\infty} g(n)h^*(n) = \frac{1}{2\pi} \int_{-\pi}^{+\pi} G(e^{j\omega})H^*(e^{j\omega}) \mathrm{d}\omega$$



$$\alpha g(n) + \beta h(n) \stackrel{DTFT}{\longleftrightarrow} \alpha G(e^{i\omega}) + \beta H(e^{i\omega})$$

$$g(-n) \stackrel{DTFT}{\longleftrightarrow} G(e^{-j\omega})$$

$$g(n - n_0) \stackrel{DTFT}{\longleftrightarrow} e^{-j\omega n_0} G(e^{i\omega})$$

$$e^{i\omega_0 n} g(n) \stackrel{DTFT}{\longleftrightarrow} G(e^{j(\omega - \omega_0)})$$

$$ng(n) \stackrel{DTFT}{\longleftrightarrow} j \frac{dG(e^{j\omega})}{d\omega}$$

$$g(n) \cdot h(n) \stackrel{DTFT}{\longleftrightarrow} \frac{1}{2\pi} \int_{-\pi}^{+\pi} G(e^{i\theta}) H(e^{i(\omega - \theta)}) d\theta$$

$$\sum_{n=-\infty}^{+\infty} g(n) h^*(n) = \frac{1}{2\pi} \int_{-\pi}^{+\pi} G(e^{i\omega}) H^*(e^{i\omega}) d\omega$$



$$\sum_{n=-\infty}^{+\infty} g(n)h^*(n) = \sum_{n=-\infty}^{+\infty} g(n) \left(\frac{1}{2\pi} \int_{-\pi}^{+\pi} H^*(e^{j\omega})e^{-j\omega n} \mathrm{d}\omega\right)$$
$$= \frac{1}{2\pi} \int_{-\pi}^{+\pi} H^*(e^{j\omega}) \left(\sum_{n=-\infty}^{+\infty} g(n)e^{-j\omega n}\right) \mathrm{d}\omega$$
$$= \frac{1}{2\pi} \int_{-\pi}^{+\pi} G(e^{j\omega})H^*(e^{j\omega}) \mathrm{d}\omega$$

Q.E.D.



• Let us compute the DTFT of the finite length exponential sequence:

$$y(n) = \begin{cases} \alpha^n & 0 \le n < M - 1 \\ 0 & \text{otherwise} \end{cases}$$

with $|\alpha| < 1$.

$$y(n) = \alpha^{n} \mu(n) - \alpha^{n} \mu(n-M) = \alpha^{n} \mu(n) - \alpha^{M} \alpha^{n-M} \mu(n-M)$$

• We already know that:

$$\alpha^{n}\mu(n) \stackrel{DTFT}{\longleftrightarrow} \frac{1}{1-\alpha e^{-j\omega}}$$

• For the linearity and the time shift properties we have

$$Y(e^{j\omega}) = \frac{1}{1 - \alpha e^{-j\omega}} - \alpha^{M} \frac{e^{-j\omega M}}{1 - \alpha e^{-j\omega}} = \frac{1 - \alpha^{M} e^{-j\omega M}}{1 - \alpha e^{-j\omega}}$$



• Let us compute the DTFT of the causal sequence v(n) defined by the finite difference equation:

$$d_0v(n) + d_1v(n-1) = p_0\delta(n) + p_1\delta(n-1)$$

with $d_0 \neq 0$, and $d_1 \neq 0$.

- Let us apply the DTFT to both terms of the finite difference equation.
- By exploiting the linearity and time-shit properties and remembering that $DTFT{\delta(n)} = 1$, we have

$$d_0 V(e^{j\omega}) + d_1 e^{-j\omega} V(e^{j\omega}) = p_0 + p_1 e^{-j\omega}$$

Thus,

$$V(e^{j\omega})=rac{p_0+p_1e^{-j\omega}}{d_0+d_1e^{-j\omega}}.$$



• Let us compute the DTFT of $y(n) = (-1)^n \alpha^n \mu(n)$ with $|\alpha| < 1$.

• It is also

$$y(n) = e^{j\pi n} \left(\alpha^n \mu(n) \right) = e^{j\pi n} x(n),$$

where we considered $x(n) = \alpha^n \mu(n)$.

• We can apply the frequency-shifting property:

$$Y(e^{j\omega}) = X(e^{j(\omega-\pi)}) = rac{1}{1-lpha e^{-j(\omega-\pi)}} = rac{1}{1+lpha e^{-j\omega}}.$$



• The total energy of a sequence g(n) is given by

$$\Xi_g = \sum_{n=-\infty}^{+\infty} |g(n)|^2.$$

• By applying the Parseval theorem with h(n) = g(n) we have

$$E_g = rac{1}{2\pi} \int_{-\pi}^{+\pi} \left| G(e^{j\omega})
ight|^2 \mathrm{d}\omega \, .$$

- $S_{gg}(e^{j\omega}) = |G(e^{j\omega})|^2$ is called **energy density spectrum**, and it defines how the energy of the sequence is distributed over the frequency spectrum.
- Let us compute the energy of the low-pass ideal signal:

$$\sum_{n=-\infty}^{+\infty} \left| h_{\rm LP}(n) \right|^2 = \frac{1}{2\pi} \int_{-\pi}^{+\pi} \left| H_{\rm LP}(e^{j\omega}) \right|^2 \mathrm{d}\omega = \frac{1}{2\pi} \int_{-\omega_c}^{+\omega_c} \mathrm{d}\omega = \frac{\omega_c}{\pi} < +\infty$$

- In general, the spectrum of a discrete-time signal is defined on the entire frequency range −π < ω ≤ +π.
 A band-limited signal has a spectrum that is limited to part of this range.
- An ideal low-pass signal has

$$X(e^{j\omega}) \left\{ egin{array}{ll}
eq 0 & 0 \leq |\omega| \leq \omega_p \ = 0 & \omega_p < |\omega| \leq \pi \end{array}
ight.$$

 ω_p is called signal *bandwidth*.

• An ideal high-pass signal has

$$X(e^{j\omega}) \left\{ egin{array}{ll} = 0 & 0 \leq |\omega| < \omega_p \
eq 0 & \omega_p \leq |\omega| \leq \pi \end{array}
ight.$$

The signal bandwidth is given by $\pi - \omega_p$.

• An ideal passband signal has

$$X(e^{j\omega}) \left\{egin{array}{ll} = 0 & 0 \leq |\omega| < \omega_a \
eq 0 & \omega_a \leq |\omega| \leq \omega_b \
eq 0 & \omega_b < |\omega| < \pi \end{array}
ight.$$

The signal bandwidth is given by $\omega_b - \omega_a$.



- We want to study the relations that link the continuous-time signals with the corresponding discrete-time signals.
- Let us consider a continuous-time signal g_a(t). We assume to uniformly sample it with sampling period T. We obtain the sequence:

$$g(n) = g_a(nT) \qquad -\infty < n < +\infty.$$

- The sampling frequency is $F_T = \frac{1}{T}$.
- The CTFT of $g_a(t)$ is

$$G_{a}(j\Omega) = \int_{-\infty}^{+\infty} g_{a}(t) e^{-j\Omega t} \mathrm{d}t,$$

and the DTFT of g(n) is

$$G(e^{j\omega}) = \sum_{n=-\infty}^{+\infty} g(n)e^{-j\omega n}$$

• We want to find the relation that exists between $G_a(j\Omega)$ and $G(e^{j\omega})$.





Mathematically, we can consider the sampling operations as the product of a signal g_a(t) and a periodic pulse train p(t), where

$$p(t) = \sum_{n=-\infty}^{+\infty} \delta(t - nT),$$

composed of an infinite sequence of Dirac pulses uniformly spaced in time and separated by the period T.

• The multiplication between $g_a(t)$ and p(t) results in a continuous-time function

$$g_p(t) = g_a(t) \cdot p(t) = \sum_{n=-\infty}^{+\infty} g_a(nT) \,\delta(t-nT),$$

which is also composed of infinite Dirac impulses, placed at time t = nT, and weighted by the sample value $g_a(nT)$.

• We will provide two expressions for the CTFT of $g_p(t)$.



- The first one is derived directly from the last expression of $g_p(n)$ considering that the CTFT of $\{\delta(t nT)\}$ is $e^{-j\Omega nT}$.
- For the linearity property of the CTFT:

$$G_p(j\Omega) = \sum_{n=-\infty}^{+\infty} g_s(nT) e^{-j\Omega nT}.$$

• By comparing this relation with the expression of $G(e^{i\omega})$, we see that

$$G(e^{j\omega}) = G_{\rho}(j\Omega)|_{\Omega = \frac{\omega}{T}}$$
$$G_{\rho}(j\Omega) = G(e^{j\omega})\Big|_{\omega = \Omega T}$$

- The DTFT $G(e^{j\omega})$ coincides with the CTFT of $g_p(n)$, apart from a frequency axis normalization.
- This normalization maps the point at $\omega = 2\pi$ of $G(e^{j\omega})$ to the point at $\Omega_T = 2\pi F_T$ of $G_p(j\Omega)$.



- In what follows, we find a second form for $G_p(j\Omega)$.
- Assume you have two continuous time signals a(t) and b(t) with CTFT A(jΩ) and B(jΩ), respectively. The CTFT of the product y(t) = a(t) · b(t) is

$$Y(j\Omega) = rac{1}{2\pi} \int_{-\infty}^{+\infty} A(j\Psi) \cdot B(j(\Omega - \Psi)) \mathrm{d}\Psi$$

 $\bullet\,$ We can use the property we have just proved to evaluate the CTFT of

$$g_p(t) = g_a(t) \cdot p(t).$$

- We assume knowing the CTFT of $g_a(t)$. Let us compute the CTFT of the pulse train p(t).
- In what follows, we prove that

$$P(j\Omega) = \frac{2\pi}{T} \sum_{k=-\infty}^{+\infty} \delta(\Omega - k \frac{2\pi}{T})$$

i.e., the CTFT of p(t) is also a uniformly spaced pulse train.





• Note that the pulse train p(t) is a periodic signal with a period of T. Thus, we can expand p(t) with the Fourier series:

$$p(t) = \sum_{k=-\infty}^{+\infty} c_k e^{j\frac{2\pi}{T}kt}$$

with

$$c_{k} = \frac{1}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} p(t) e^{-j\frac{2\pi}{T}kt} dt = \frac{1}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} \delta(t) e^{-j\frac{2\pi}{T}kt} dt = \frac{1}{T} e^{-j\frac{2\pi}{T}k0} = \frac{1}{T}$$

• Thus,

$$p(t) = \frac{1}{T} \sum_{k=-\infty}^{+\infty} e^{j\frac{2\pi}{T}kt}$$

and for the linearity of the CTFT

$$P(j\Omega) = \frac{1}{T} \sum_{k=-\infty}^{+\infty} \mathsf{CTFT}\left\{e^{j\frac{2\pi}{T}kt}\right\}.$$





• We can easily verify that

$$\mathsf{CTFT}\left\{e^{j\frac{2\pi}{T}kt}\right\} = 2\pi\delta(\Omega - \frac{2\pi}{T}k)$$

• Indeed, the IDFT of the right-hand side is

$$\frac{1}{2\pi}\int_{-\infty}^{+\infty}2\pi\delta(\Omega-\frac{2\pi}{T}k)e^{j\Omega t}\mathrm{d}\Omega=e^{j\frac{2\pi}{T}kt}$$

• Replacing CTFT $\left\{e^{j\frac{2\pi}{T}kt}\right\}$ in the expression of $P(j\Omega)$, we have

$$P(j\Omega) = \frac{2\pi}{T} \sum_{k=-\infty}^{+\infty} \delta(\Omega - k \frac{2\pi}{T}).$$

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• We can now compute $G_p(j\Omega)$

$$G_{p}(j\Omega) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} G_{a}(j\Psi) \cdot P(j(\Omega - \Psi)) d\Psi =$$

$$= \frac{1}{2\pi} \int_{-\infty}^{+\infty} G_{a}(j\Psi) \cdot \frac{2\pi}{T} \sum_{k=-\infty}^{+\infty} \delta(\Omega - \Psi - k\frac{2\pi}{T}) d\Psi =$$

$$= \frac{1}{T} \sum_{k=-\infty}^{+\infty} \int_{-\infty}^{+\infty} G_{a}(j\Psi) \cdot \delta(\Omega - \Psi - k\frac{2\pi}{T}) d\Psi =$$

$$= \frac{1}{T} \sum_{k=-\infty}^{+\infty} G_{a}(j(\Omega - k\frac{2\pi}{T})) =$$

$$= \frac{1}{T} \sum_{k=-\infty}^{+\infty} G_{a}(j(\Omega - k\Omega_{T}))$$

with $\Omega_T = \frac{2\pi}{T}$.



• Consequently,

$$G(e^{j\omega}) = G_{\rho}(j\Omega)|_{\Omega = \frac{\omega}{T}} = \frac{1}{T} \sum_{k=-\infty}^{+\infty} G_{\mathfrak{s}} \big[j \big(\frac{\omega}{T} + k\Omega_T \big) \big] = \frac{1}{T} \sum_{k=-\infty}^{+\infty} G_{\mathfrak{s}} \big[j \big(\frac{\omega + 2\pi k}{T} \big) \big].$$

- The DTFT is given by the periodic repetition, with a period of 2π , of the continuous spectrum $G_a\left(j\frac{\omega}{T}\right)$.
- $G_a\left(j\frac{\omega}{T}\right)$ is identical to $G_a\left(j\Omega\right)$, but the frequency axis has been normalized such that $\omega = 2\pi$ corresponds to the angular sampling frequency $\Omega_T = \frac{2\pi}{T}$.
- In order to avoid any overlap between the repeated spectra, it must be

$$G_{a}\left(j\frac{\omega}{T}\right) = 0$$
 for $|\omega| > \pi$,

or, equivalently,

$$G_{a}\left(j\Omega
ight)=0 \hspace{1cm} ext{for} \hspace{1cm} |\Omega|>rac{\pi}{T}=rac{\Omega_{T}}{2}.$$

- If this condition is satisfied, the discrete spectrum reproduces in the band [-π, +π] the continuous spectrum.
- If this condition is not satisfied, we have a distortion caused by the overlap of the tails of the spectrum. This distortion is called **aliasing**.

Digital Signal and Image Processing





- The frequency $\frac{1}{2T} = \frac{F_T}{2}$ is called **Nyquist frequency**.
- Similarly, the angular frequency $\frac{\Omega_T}{2}$ is called **Nyquist** angular frequency.
- Let us assume, for example, that our signal occupies the band $[-\Omega_m, +\Omega_m].$
- If $\Omega_T \ge 2\Omega_m$, the different repetitions of the spectrum do not overlap.
- If Ω_T < 2Ω_m, the repetitions of the spectrum overlap, and we have an aliasing error.

The sampling theorem

- If the continuous-time signal spectrum is preserved in the discrete domain (i.e., if we do not have aliasing), we can reconstruct the original signal from the samples $g(n) = g_a(nT)$.
- For this purpose, let us build the signal:

$$g_p(t) = \sum_{n=-\infty}^{+\infty} g(n)\delta(t-nT) = \sum_{n=-\infty}^{+\infty} g_a(nT)\delta(t-nT)$$

• We know its CTFT:

$$G_p(j\Omega) = rac{1}{T}\sum_{k=-\infty}^{+\infty}G_a\left[j(\Omega+k\Omega_T)
ight].$$

This continuous spectrum is given by the periodic repetition of the spectrum $G_a(j\Omega)$ with period Ω_T .

- If the signal has been sampled with a frequency $\Omega_T > 2\Omega_m$ we do not have aliasing, and the repetitions of $G_a(j\Omega)$ do not overlap.
- Thus, we can faithfully reconstructg_a(t) by passing g_p(t) through a filter that lets the spectrum in band
 [-Ω_m, Ω_m] pass without any alteration, while stopping all signal components outside that frequency range.
- We will see that such a filter is an ideal low-pass filter with bandwidth Ω_m .
- We have just proved the theorem that is at the basis of all signal processing and modern telecommunications, the sampling theorem.

Digital Signal and Image Processing





Let $g_a(t)$ be a band-limited signal, with $G_a(j\Omega) = 0$ for $|\Omega| > \Omega_m$. Then, $g_a(t)$ is uniquely determined by its samples $g_a(nT)$, (i.e., can be faithfully reconstructed by its samples $g_a(nT)$,) $-\infty < n < +\infty$, if the angular sampling frequency $\Omega_T > 2\Omega_m$

with $\Omega_T = rac{2\pi}{T}$.

- In other words, if we want to be able to recover the band-limited signal $g_a(t)$ from its samples, we must sample the signal with a frequency at least twice the signal bandwidth.
- The signal can be recovered by generating the signal

$$g_p(t) = \sum_{n=-\infty}^{+\infty} g(n)\delta(t-nT)$$

and by filtering this signal with an ideal low-pass filter.



- In general, real-world signals are not band-limited. They occupy an infinite band, but most of their energy is concentrated at the low frequencies.
- To avoid distortions due to aliasing, the signal is typically filtered with a low-pass filter before sampling.
- This filter cuts off high frequencies and generates a band-limited signal. Such a filter is called an **anti-aliasing** filter.
- After filtering, the signal can be safely sampled with a frequency greater than $2\Omega_m$ and processed as desired.



Figure 3.12: Block diagram representation of the discrete-time digital processing of a continuous-time signal.

- Also the signal that comes out of the Digital to Analog (D/A) converter is filtered with a low-pass filter, called the **reconstruction filter** or **anti-imaging filter**.
- In this way, all the frequencies (the images) outside the band of the original signal, $[-\Omega_m, +\Omega_m]$ are removed.



- For more information study:
 - S. K. Mitra, "Digital Signal Processing: a computer based approach," 4th edition, McGraw-Hill, 2011

Chapter 3.1, pp. 89-93 Chapter 3.2-3.5, pp. 94-112 Chapter 3.8, pp. 115-124

Unless otherwise specified, all images have either been originally produced or have been taken from S. K. Mitra, "Digital Signal Processing: a computer based approach."