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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

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

• 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.

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- 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

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

- 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

$$
x_a(t) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} X_a(j\Omega) e^{i\Omega t} d\Omega.
$$

• 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:

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

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

- \bullet $\big|X_a(j\Omega)\big|$ 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.,

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

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

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

$$
\int_{-\infty}^{+\infty} |x_a(t)| = \int_0^{+\infty} e^{-\alpha t} dt = -\left. \frac{e^{-\alpha t}}{\alpha} \right|_0^{+\infty} = 0 - \left(-\frac{1}{\alpha} \right) = \frac{1}{\alpha}
$$

CTFT Example

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

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:

$$
\delta(t) = \begin{cases} 0 & t \neq 0 \\ +\infty & t = 0 \end{cases}
$$

with

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

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

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

• 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

 $X_a(i\Omega) = \alpha F_a(i\Omega) + \beta G_a(i\Omega).$

• 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

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

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

$$
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$$

- An absolutely integrable signal, i.e., a signal for which $\int_{-\infty}^{+\infty}$ $|x_a(t)|dt < +\infty$, 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:

$$
E_x = \int_{-\infty}^{+\infty} |x_a(t)|^2 dt = \frac{1}{2\pi} \int_{-\infty}^{+\infty} |X_a(j\Omega)|^2 d\Omega
$$

- \bullet $S_{xx}(\Omega)=\big|X_a(j\Omega)\big|^2$ 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 $\Omega_a \leq \Omega \leq \Omega_b$ can be computed by integrating $S_\infty(\Omega)$ over the interval $[\Omega_a, \Omega_b]$:

$$
E_{x,r} = \frac{1}{2\pi} \int_{\Omega_a}^{\Omega_b} S_{xx}(\Omega) d\Omega.
$$

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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 $-\infty < \Omega < +\infty$.
- \bullet An **ideal band-limited signal** has spectrum that is zero outside a certain range Ω _a ≤ $|\Omega|$ ≤ Ω _b, with $0 \leq \Omega_a, \Omega_b \leq +\infty$, i.e.,

$$
X_a(j\Omega)=\left\{\begin{array}{ll}0 & \qquad 0\leq|\Omega|<\Omega_a \\ 0 & \qquad \Omega_b<|\Omega|<+\infty \end{array}\right.
$$

- 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 $[\Omega_a, \Omega_b]$ of interest.
- A signal is called low-pass if its spectrum occupies the frequency range $0 < |\Omega| < \Omega$, where Ω_p is called the signal bandwidth.
- A signal is called **high-pass** if its spectrum occupies the frequency range $\Omega_p \leq |\Omega| \leq +\infty$, and the signal bandwidth extends from Ω_p to ∞ .
- Eventually, a signal is called passband if its spectrum occupies the frequency range $0 < \Omega_L \leq |\Omega| \leq \Omega_H < +\infty$, where $\Omega_H - \Omega_L$ is the signal bandwidth.

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- 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 $e^{j\omega n}$, where ω is the *normalized angular frequency*, and $\omega\in\mathbb{R}.$
- 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

 $\bullet\,$ Let us compute the DTFT of the causal exponential sequence $x(n)=\alpha^n\mu(n)$, where $|\alpha|< 1$ and $\mu(n)$ is the unit step sequence.

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

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

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

$$
\sum_{n=0}^{+\infty} q^n = \frac{1}{1-q} \qquad \text{if } |q| < 1.
$$

since

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

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

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

- Note that $\sum_{n=1}^{+\infty} x(n) e^{-j\omega n}$ is the Fourier series expansion of the periodic function $X(e^{j\omega})$. $n=-\infty$
- \bullet The coefficients of the Fourier series, $x(n)$, can be computed from $X(e^{j\omega})$ using the Fourier integral:

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

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 ω weighted by the complex function $X(e^{j\omega})$.

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IDTFT proof

• Let us verify that

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

 \bullet 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 $n \neq l$,

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

• If $n = l$,

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

• Thus, in general, it is

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

ü

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

• Eventually, we have

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

Q.E.D.

A. Carini **Example 20** 20 / 49

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

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

with $\mathcal{X}_\mathrm{re}(\mathrm{e}^{j\omega})$ and $\mathcal{X}_\mathrm{im}(\mathrm{e}^{j\omega})\in\mathbb{R}$.

$$
X_{\rm re}(e^{j\omega}) = \frac{1}{2} \left[X(e^{j\omega}) + X^*(e^{j\omega}) \right],
$$

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

• The DTFT can be expressed in polar form as

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

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

$$
\theta(\omega) = \arctan \frac{X_{\text{im}}(e^{j\omega})}{X_{\text{re}}(e^{j\omega})}
$$

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

A. Carini **Example 21** 21 / 49 and 21 an

• 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
	- $X_{\rm re}(\mathrm{e}^{-j\omega})=X_{\rm re}(\mathrm{e}^{j\omega})$ is an even function, $\mathcal{X}_\mathrm{im}(\mathrm{e}^{-j\omega}) = -\mathcal{X}_\mathrm{im}(\mathrm{e}^{j\omega})$ 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 $(x(n) = x(-n))$,

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

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

 $X(e^{j\omega})=jX_{\mathrm{im}}(e^{j\omega})$ 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:

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

.

A. Carini **Example 23** / 49 and 23 / 49 and 23 and 23

• 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:

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

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

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

• For $n = 0$

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

• Thus,

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

• 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

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

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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:

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

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

$$
\chi(n) = \frac{1}{2\pi} 2\pi \int_{\text{Digital Sign}}^{+\pi} \delta(\omega - \omega_0) e^{j\omega n} \mathrm{d}\omega = e^{j\omega_0 n}
$$
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- \bullet 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 $j\frac{\mathrm{d} G(e^{j\omega})}{n\omega}$ $\frac{1}{d\omega}$.
- Modulation: The DTFT of $g(n) \cdot h(n)$ is $\frac{1}{2\pi}$ $\int^{+\pi}$ $\int_{-\pi}^{\pi} G(e^{j\theta}) H(e^{j(\omega-\theta)}) d\theta.$
- Parseval theorem:

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

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

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

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

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

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

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

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

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

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

$$
= \frac{1}{2\pi} \int_{-\pi}^{+\pi} G(e^{j\omega}) H^*(e^{j\omega}) d\omega
$$

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

$$
d_0V(e^{j\omega})+d_1e^{-j\omega}V(e^{j\omega})=p_0+p_1e^{-j\omega}
$$

Thus,

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

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

• It is also

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

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

• We can apply the frequency-shifting property:

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

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

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

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

$$
E_g = \frac{1}{2\pi} \int_{-\pi}^{+\pi} \left| G(e^{j\omega}) \right|^2 d\omega.
$$

- \bullet $S_{gg}(e^{j\omega}) = |G(e^{j\omega})|^2$ 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:

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

Band-limited signals

- In general, the spectrum of a discrete-time signal is defined on the entire frequency range $-\pi < \omega \leq +\pi$. A band-limited signal has a spectrum that is limited to part of this range.
- An ideal low-pass signal has

$$
X(e^{j\omega})\begin{cases}\neq 0 & 0\leq |\omega|\leq \omega_p\\=0 & \omega_p<|\omega|\leq \pi\end{cases}
$$

 ω_p is called signal bandwidth.

• An ideal high-pass signal has

$$
X(e^{j\omega})\begin{cases}=0 & 0 \leq |\omega| < \omega_p\\ \neq 0 & \omega_p \leq |\omega| \leq \pi\end{cases}
$$

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

• An ideal passband signal has

$$
X(e^{j\omega})\begin{cases}\n=0 & 0 \leq |\omega| < \omega_a \\
\neq 0 & \omega_a \leq |\omega| \leq \omega_b \\
=0 & \omega_b < |\omega| < \pi\n\end{cases}
$$

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:

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

- \bullet The sampling frequency is $F_{\mathcal{T}}=\frac{1}{\mathcal{T}}$ $\frac{1}{T}$.
- 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

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

.

 \bullet We want to find the relation that exists between $\mathit{G}_{a}(j\Omega)$ and $\mathit{G}(e^{j\omega}).$

• 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:

$$
G_p(j\Omega)=\sum_{n=-\infty}^{+\infty}g_a(n\mathcal{T})e^{-j\Omega n\mathcal{T}}.
$$

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

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

$$
G_p(j\Omega) = G(e^{j\omega})|_{\omega = \Omega T}
$$

- 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 $A(j\Omega)$ and $B(j\Omega)$, respectively. The CTFT of the product $y(t) = a(t) \cdot b(t)$ is

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

• 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.

Proof

• 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

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

• Thus,

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

and for the linearity of the CTFT

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

• We can easily verify that

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

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

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

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

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

• We can now compute $G_p(j\Omega)$

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

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

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

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

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

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

• Consequently,

$$
G(e^{j\omega}) = G_p(j\Omega)|_{\Omega = \frac{\omega}{\tau}} = \frac{1}{\tau} \sum_{k=-\infty}^{+\infty} G_a[j(\frac{\omega}{\tau} + k\Omega_{\tau})] = \frac{1}{\tau} \sum_{k=-\infty}^{+\infty} G_a[j(\frac{\omega + 2\pi k}{\tau})].
$$

- The DTFT is given by the periodic repetition, with a period of 2π , of the continuous spectrum $G_a\left(j\frac{\omega}{\tau}\right)$ T .
- $G_a \left(j \frac{\omega}{7} \right)$ T) is identical to $\mathcal{G}_a(j\Omega)$, 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

$$
G_a\left(j\frac{\omega}{T}\right)=0 \quad \text{ for } \quad |\omega|>\pi,
$$

or, equivalently,

$$
G_a(j\Omega)=0 \quad \text{for} \quad |\Omega| > \frac{\pi}{T} = \frac{\Omega_T}{2}.
$$

- If this condition is satisfied, the discrete spectrum reproduces in the band $[-\pi, +\pi]$ 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.

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- $AX(SQ)$ \mathcal{Q}_{n} $\overline{\Omega}$ -0 $280 - 25 > 28$ $\{X(G_2)\}$ w $+$ $X(e^{iv})$ -22 28 $40U$ Ω = 2 Ω $\boxed{\pm\rho}$ $\mathbb{Z}(\mathfrak{i}\frac{\omega}{\mathfrak{p}})$ $+27$ $X(e^{3w})$ alissin error
- The frequency $\frac{1}{2T} = \frac{F_T}{2}$ $\frac{1}{2}$ 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 $\Omega_{\tau} \geq 2\Omega_m$, the different repetitions of the spectrum do not overlap.
- If Ω _T $<$ 2 Ω _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:

$$
G_{\rho}(j\Omega)=\frac{1}{T}\sum_{k=-\infty}^{+\infty}\, G_{a}\left[j(\Omega+k\Omega_{T})\right].
$$

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 $G_a(i\Omega)$ do not overlap.
- Thus, we can faithfully reconstruct $g_a(t)$ by passing $g_p(t)$ through a filter that lets the spectrum in band $[-\Omega_m, \Omega_m]$ 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.

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

$$
\Omega_{\mathcal{T}}\geq 2\Omega_m
$$

with $\Omega_{\mathcal{T}} = \frac{2\pi}{\mathcal{T}}$ $\frac{1}{T}$.

- 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

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

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:
	- F 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."