

# **Discrete Time Fourier Series (DTFS) and Fourier Transform (DTFT)**

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## Recall: Four Classes of Fourier Representations

Time Property	Periodic	Nonperiodic
Continuous (t)	Fourier Series (FS) [Chapter 3.5]	Fourier Transform (FT) [Chapter 3.7]
Discrete (n)	Discrete Time Fourier Series (DTFS) [Chapter 3.4]	Discrete Time Fourier Transform (DTFT) [Chapter 3.6]

## Recall: Frequency Response of DT Systems

Consider a discrete-time LTI system with impulse response  $h[n]$ .

- Suppose a **complex sinusoid** is applied as input. i.e.,  $x[n] = e^{j\Omega n}$ .
- Then, the output  $y[n]$  of the DT LTI system is given by

$$\begin{aligned} y[n] &= \sum_{k=-\infty}^{\infty} h[k]x[n-k] = \sum_{k=-\infty}^{\infty} h[k]e^{j\Omega(n-k)} \\ &= e^{j\Omega n} \sum_{k=-\infty}^{\infty} h[k]e^{-j\Omega k} = H(e^{j\Omega})e^{j\Omega n}, \quad \text{where} \end{aligned}$$

$$\boxed{H(e^{j\Omega}) = \sum_{k=-\infty}^{\infty} h[k]e^{-j\Omega k}} = \text{Frequency Response.}$$

## Recall: Frequency Response of DT Systems

Substituting  $H(e^{j\Omega}) = |H(e^{j\Omega})|e^{j \arg\{H(e^{j\Omega})\}}$ , we obtain the output of the DT LTI system corresponding to a complex sinusoid  $e^{j\Omega n}$  as

$$y[n] = H(e^{j\Omega})e^{j\Omega n} = |H(e^{j\Omega})|e^{j(\Omega n + \arg\{H(e^{j\Omega})\})},$$

which implies that **in steady state** the DT LTI system modifies the magnitude of the input by a factor  $|H(e^{j\Omega})|$  and modifies the phase by a shift of  $\arg\{H(e^{j\Omega})\}$ .

Therefore,  $|H(e^{j\Omega})|$  is called the **magnitude response** and  $\arg\{H(e^{j\Omega})\}$  is called the **phase response** of the DT LTI system.

# Discrete-Time Fourier Series (DTFS) for Periodic Signals

## Harmonically-Related DT Complex Sinusoids

- A set of periodic complex sinusoids is said to be **harmonically related** if there exists some constant  $2\pi/N$  such that the fundamental frequency of each complex sinusoid is an integer multiple of  $2\pi/N$ .

- Consider the set of harmonically-related complex sinusoids given by

$$\phi_k(n) = e^{j(2\pi/N)kn} \quad \text{for all integer } k.$$

- In the above set  $\{\phi_k\}$ , only  $N$  elements are distinct, since

$$\phi_k = \phi_{k+N} \quad \text{for all integer } k.$$

- Since the fundamental frequency of each of the harmonically-related complex sinusoids is an integer multiple of  $\frac{2\pi}{N}$ , a linear combination of these complex sinusoids must be  $N$ -periodic.

## DTFS Definition

- A periodic complex-valued sequence  $x$  with fundamental period  $N$  can be represented as a linear combination of harmonically-related complex sinusoids as

$$x(n) = \sum_{k=\langle N \rangle} a_k e^{j(2\pi/N)kn},$$

where  $\sum_{k=\langle N \rangle}$  denotes summation over any  $N$  consecutive integers (e.g.,  $0, 1, \dots, N-1$ ). (The summation can be taken over any  $N$  consecutive integers, due to the  $N$ -periodic nature of  $x$  and  $e^{j(2\pi/N)kn}$ .)

- The above representation of  $x$  is known as the (DT) **Fourier series** and the  $a_k$  are called **Fourier series coefficients**.

- **Notation:** Here,  $\boxed{\Omega_0 = (2\pi/N)}$  denotes the **fundamental (angular) frequency** of the signal  $x[n]$ , and  $\boxed{\Omega_k = k\Omega_0 = (2k\pi/N)}$  denotes its **k-th harmonic frequency**.

## DTFS Definition

- A periodic sequence  $x$  with fundamental period  $N$  has the Fourier series coefficient sequence  $a$  given by

$$a_k = \frac{1}{N} \sum_{n=\langle N \rangle} x(n) e^{-j(2\pi/N)kn}.$$

(The summation can be taken over any  $N$  consecutive integers due to the  $N$ -periodic nature of  $x$  and  $e^{-j(2\pi/N)kn}$ .)

Notation:

- Here,  $\boxed{a_k = X[k]}$ ,  $k = 0, \dots, N - 1$ , denote the DTFS coefficients.
- Since both  $x[n]$  and  $e^{j(2\pi/N)kn} = e^{j\Omega_0 kn}$  are periodic functions with period  $N$ , the DTFS coefficients  $a_k$  or  $X[k]$ , as a function of index  $k$ , is also periodic with period  $N$ .



## DTFS Definition

- To denote that the DT signal (or, sequence)  $x[n]$  has the Fourier series coefficient sequence  $X[k]$ , we write

$$x[n] \overset{\text{DTFS}}{\longleftrightarrow} X[k].$$

- The DTFS is the only Fourier representation that can be numerically evaluated and manipulated in a computer.
- This is because both the time and the frequency domain representations of the signal are exactly characterized by a finite set of  $N$  numbers.
- DTFS is used for approximating the other three Fourier representations for the purpose of implementation on a computer.

## DTFS Examples and Exercises

Example: (Method of Inspection) Determine the DTFS coefficients of  $x[n] = \cos(\pi n/3 + \phi)$ . Sketch the magnitude and phase spectrum.

*Solution: The period is  $N = 6$ . The fundamental frequency is  $\Omega_0 = 2\pi/N = \pi/3$ . The DTFS coefficients are:*

$$X[1] = e^{j\phi}/2, \quad X[-1] = e^{-j\phi}/2.$$

*Can you draw the magnitude and phase plot?*

Exercise: Determine the DTFS coefficients of  $x[n] = 1 + \sin(\pi n/12 + 3\pi/8)$ . Sketch the magnitude and phase spectrum.

## DTFS Examples and Exercises

**Example:** Determine the DTFS coefficients of the  $N$ -periodic impulse train given by

$$x[n] = \sum_{\ell=-\infty}^{\infty} \delta[n - \ell N].$$

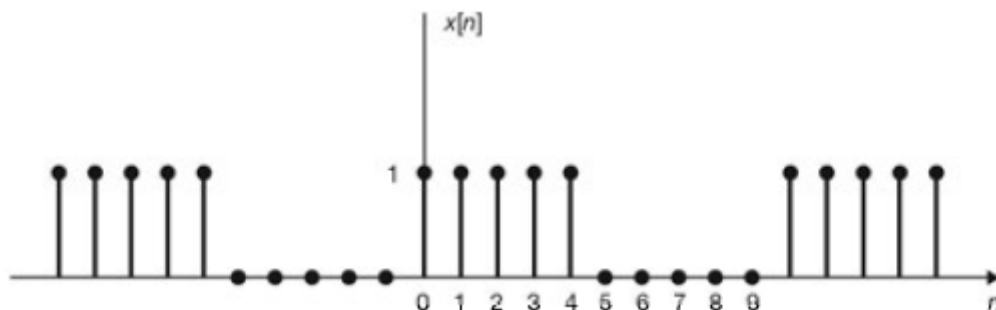
Sketch the magnitude and phase spectrum.

*Solution:* We have

$$X[k] = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-jkn2\pi/N} = \frac{1}{N} \sum_{n=0}^{N-1} \delta[n] e^{-jkn2\pi/N} = \frac{1}{N}.$$

## DTFS Examples and Exercises

**Example:** Determine the DTFS coefficients of the signal shown below. Sketch the magnitude and phase spectrum.



**Solution:** We have  $N = 10$  and  $\Omega_0 = 2\pi/N = \pi/5$ . Therefore, the DTFS coefficients are given by

## DTFS Examples and Exercises

*Solution: (continued..)*

$$\begin{aligned} X[k] &= \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-jkn2\pi/N} = \frac{1}{10} \sum_{n=0}^4 e^{-jkn\pi/5} \\ &= \frac{1}{10} \frac{1 - e^{-jk\pi}}{1 - e^{-jk(\pi/5)}} = \frac{1}{10} \frac{e^{-jk(\pi/2)}}{e^{-jk(\pi/10)}} \frac{e^{jk(\pi/2)} - e^{-jk(\pi/2)}}{e^{jk(\pi/10)} - e^{-jk(\pi/10)}}. \\ &= \frac{1}{10} e^{-jk(2\pi/5)} \frac{\sin(k(\pi/2))}{\sin(k(\pi/10))}, \quad k = 0, 1, \dots, 9. \end{aligned}$$

## DTFS Trigonometric Form

Suppose  $N$  is even; then, we can rewrite the DTFS by separating the terms corresponding to  $k = 0$  and  $k = N/2$  from the remaining terms as

$$\begin{aligned}x[n] &= \sum_{k=-(N/2)+1}^{N/2} X[k]e^{jk\Omega_0 n} \\&= X[0] + X[N/2]e^{j\pi n} + \sum_{k=1}^{(N/2)-1} X[k]e^{jk\Omega_0 n} + X[-k]e^{-jk\Omega_0 n} \\&= X[0] + X[N/2] \cos(\pi n) + \sum_{k=1}^{(N/2)-1} \{ B[k] \cos(k\Omega_0 n) + A[k] \sin(k\Omega_0 n) \},\end{aligned}$$

where  $\boxed{B[k] = X[k] + X[-k]}$  and  $\boxed{A[k] = j(X[k] - X[-k])}$ .

## DTFS Trigonometric Form

Suppose  $N$  is odd; then, we can rewrite the DTFS by separating the term corresponding to  $k = 0$  from the remaining terms as

$$\begin{aligned}x[n] &= \sum_{k=-(N-1)/2}^{(N-1)/2} X[k] e^{jk\Omega_0 n} \\&= X[0] + \sum_{k=1}^{(N-1)/2} X[k] e^{jk\Omega_0 n} + X[-k] e^{-jk\Omega_0 n} \\&= X[0] + \sum_{k=1}^{(N-1)/2} \{ B[k] \cos(k\Omega_0 n) + A[k] \sin(k\Omega_0 n) \},\end{aligned}$$

where  $\boxed{B[k] = X[k] + X[-k]}$  and  $\boxed{A[k] = j(X[k] - X[-k])}$ .

## DTFS Properties



## Linearity

- Let  $x$  and  $y$  be  $N$ -periodic sequences. If  $x(n) \xleftrightarrow{\text{DTFS}} a_k$  and  $y(n) \xleftrightarrow{\text{DTFS}} b_k$ , then

$$\alpha x(n) + \beta y(n) \xleftrightarrow{\text{DTFS}} \alpha a_k + \beta b_k,$$

where  $\alpha$  and  $\beta$  are complex constants.

- That is, a linear combination of sequences produces the same linear combination of their Fourier series coefficients.

## Translation (Time Shifting)

- Let  $x$  denote a periodic sequence with period  $N$ . If  $x(n) \xleftrightarrow{\text{DTFS}} c_k$ , then

$$x(n - n_0) \xleftrightarrow{\text{DTFS}} e^{-jk(2\pi/N)n_0} c_k,$$

where  $n_0$  is an integer constant.

- In other words, time shifting a periodic sequence changes the argument (but not magnitude) of its Fourier series coefficients.

## Modulation (Frequency Shifting)

- Let  $x$  denote a periodic sequence with period  $N$ . If  $x(n) \xleftrightarrow{\text{DTFS}} c_k$ , then

$$e^{j(2\pi/N)k_0n}x(n) \xleftrightarrow{\text{DTFS}} c_{k-k_0},$$

where  $k_0$  is an integer constant.

- That is, multiplying a sequence by a complex sinusoid whose frequency is an integer multiple of  $2\pi/N$  results in a translation of the corresponding Fourier series coefficient sequence.

## Reflection (Time Reversal)

- Let  $x$  denote a periodic sequence with period  $N$ . If  $x(n) \xleftrightarrow{\text{DTFS}} c_k$ , then

$$x(-n) \xleftrightarrow{\text{DTFS}} c_{-k}.$$

- That is, time reversing a sequence results in a time reversal of the corresponding Fourier series coefficient sequence.

## Conjugation

- Let  $x$  denote a periodic sequence with period  $N$ . If  $x(n) \xleftrightarrow{\text{DTFS}} c_k$ , then

$$x^*(n) \xleftrightarrow{\text{DTFS}} c_{-k}^*.$$

- In other words, conjugating a sequence has the effect of time reversing and conjugating the corresponding Fourier series coefficient sequence.

## Duality

- Let  $x$  denote a periodic sequence with period  $N$ . If  $x(n) \xleftrightarrow{\text{DTFS}} a(k)$ , then

$$a(n) \xleftrightarrow{\text{DTFS}} \frac{1}{N}x(-k).$$

- This is known as the **duality property** of Fourier series.
- This property follows from the high degree of symmetry in the analysis and synthesis Fourier-series equations, which are respectively given by

$$x(m) = \sum_{\ell=\langle N \rangle} a(\ell)e^{j(2\pi/N)\ell m} \quad \text{and} \quad a(m) = \frac{1}{N} \sum_{\ell=\langle N \rangle} x(\ell)e^{-j(2\pi/N)m\ell}.$$

- That is, the analysis and synthesis equations are identical except for a *factor of  $N$*  and *different sign* in the parameter for the exponential function.

## Periodic Convolution

- Let  $x$  and  $y$  be  $N$ -periodic sequences. If  $x(n) \xleftrightarrow{\text{DTFS}} a_k$  and  $y(n) \xleftrightarrow{\text{DTFS}} b_k$ , then

$$x \circledast y(n) \xleftrightarrow{\text{DTFS}} N a_k b_k.$$

- That is, periodic convolution of two sequences multiplies their corresponding Fourier series coefficient sequences (up to a scale factor).

**Notation:** Here, the **circular convolution** of  $x[n]$  and  $y[n]$  is given by

$$x[n] \circledast h[n] = \sum_{m=0}^{N-1} x[m]y[n - m].$$

## Multiplication

- Let  $x$  and  $y$  be  $N$ -periodic sequences. If  $x(n) \xleftrightarrow{\text{DTFS}} a_k$  and  $y(n) \xleftrightarrow{\text{DTFS}} b_k$ , then

$$x(n)y(n) \xleftrightarrow{\text{DTFS}} a \circledast b(k).$$

- That is, multiplying two sequences results in a circular convolution of their corresponding Fourier series coefficient sequences.

**Notation:** Here, the **circular convolution** of  $a_k = X[k]$  and  $b_k = Y[k]$  is given by

$$X[k] \circledast Y[k] = \sum_{m=0}^{N-1} X[m]Y[k - m].$$



## Parseval's relation

- A sequence  $x$  and its Fourier series coefficient sequence  $a$  satisfy the following relationship:

$$\frac{1}{N} \sum_{n=\langle N \rangle} |x(n)|^2 = \sum_{k=\langle N \rangle} |a_k|^2.$$

- The above relationship is simply stating that the amount of energy in a single period of  $x$  and the amount of energy in a single period of  $a$  are equal up to a scale factor.
- In other words, the transformation between a sequence and its Fourier series coefficient sequence preserves energy (up to a scale factor).

## Other Properties

- For an  $N$ -periodic sequence  $x$  with Fourier-series coefficient sequence  $a$ , the following properties hold:

$x$  is even  $\Leftrightarrow a$  is even;    and

$x$  is odd  $\Leftrightarrow a$  is odd.

- In other words, the even/odd symmetry properties of  $x$  and  $a$  always match.

## Other Properties

- A sequence  $x$  is *real* if and only if its Fourier series coefficient sequence  $a$  satisfies

$$a_k = a_{-k}^* \text{ for all } k$$

(i.e.,  $a$  is *conjugate symmetric*).

- From properties of complex numbers, one can show that  $a_k = a_{-k}^*$  is equivalent to

$$|a_k| = |a_{-k}| \quad \text{and} \quad \arg a_k = -\arg a_{-k}$$

(i.e.,  $|a_k|$  is *even* and  $\arg a_k$  is *odd*).

- Note that  $x$  being real does *not* necessarily imply that  $a$  is real.

## Other Properties

- For an  $N$ -periodic sequence  $x$  with Fourier-series coefficient sequence  $a$ , the following properties hold:
  - 1  $a_0$  is the average value of  $x$  over a single period;
  - 2  $x$  is real and even  $\Leftrightarrow a$  is real and even; and
  - 3  $x$  is real and odd  $\Leftrightarrow a$  is purely imaginary and odd.

$$x(n) \xleftrightarrow{\text{DTFS}} a_k \quad \text{and} \quad y(n) \xleftrightarrow{\text{DTFS}} b_k$$

Property	Time Domain	Fourier Domain
Linearity	$\alpha x(n) + \beta y(n)$	$\alpha a_k + \beta b_k$
Translation	$x(n - n_0)$	$e^{-jk(2\pi/N)n_0} a_k$
Modulation	$e^{j(2\pi/N)k_0 n} x(n)$	$a_{k-k_0}$
Reflection	$x(-n)$	$a_{-k}$
Conjugation	$x^*(n)$	$a_{-k}^*$
Duality	$a_n$	$\frac{1}{N} x(-k)$
Periodic Convolution	$x \circledast y(n)$	$N a_k b_k$
Multiplication	$x(n)y(n)$	$a \circledast b_k$

Property	
Parseval's Relation	$\frac{1}{N} \sum_{n=\langle N \rangle}  x(n) ^2 = \sum_{k=\langle N \rangle}  a_k ^2$
Even Symmetry	$x \text{ is even} \Leftrightarrow a \text{ is even}$
Odd Symmetry	$x \text{ is odd} \Leftrightarrow a \text{ is odd}$
Real / Conjugate Symmetry	$x \text{ is real} \Leftrightarrow a \text{ is conjugate symmetric}$



# Discrete Time Fourier Transform (DTFT)

## DTFT from DTFS

- Recall that the Fourier series representation of a  $N$ -periodic sequence  $x$  is given by

$$x(n) = \sum_{k=\langle N \rangle} \underbrace{\left( \frac{1}{N} \sum_{\ell=\langle N \rangle} x(\ell) e^{-j(2\pi/N)k\ell} \right)}_{c_k} e^{j(2\pi/N)kn}.$$

- In the above representation, if we take the limit as  $N \rightarrow \infty$ , we obtain

$$x(n) = \frac{1}{2\pi} \int_{2\pi} \underbrace{\left( \sum_{n=-\infty}^{\infty} x(n) e^{-j\Omega n} \right)}_{X(\Omega)} e^{j\Omega n} d\Omega$$



## FT and Inverse FT Equations

- The **Fourier transform** of the sequence  $x$ , denoted  $\mathcal{F}x$  or  $X$ , is given by

$$\mathcal{F}x(\Omega) = X(\Omega) = \sum_{n=-\infty}^{\infty} x(n)e^{-j\Omega n}.$$

- The preceding equation is sometimes referred to as **Fourier transform analysis equation** (or **forward Fourier transform equation**).
- The **inverse Fourier transform** of  $X$ , denoted  $\mathcal{F}^{-1}X$  or  $x$ , is given by

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

- The preceding equation is sometimes referred to as the **Fourier transform synthesis equation** (or **inverse Fourier transform equation**).

## DTFT Convergence

- For a sequence  $x$ , the Fourier transform analysis equation (i.e.,  $X(\Omega) = \sum_{-\infty}^{\infty} x(n)e^{-j\Omega n}$ ) converges *uniformly* if

$$\sum_{k=-\infty}^{\infty} |x(k)| < \infty$$

(i.e.,  $x$  is *absolutely summable*).

- For a sequence  $x$ , the Fourier transform analysis equation (i.e.,  $X(\Omega) = \sum_{-\infty}^{\infty} x(n)e^{-j\Omega n}$ ) converges in the *MSE sense* if

$$\sum_{k=-\infty}^{\infty} |x(k)|^2 < \infty$$

(i.e.,  $x$  is *square summable*).

- For a bounded Fourier transform  $X$ , the Fourier transform synthesis equation (i.e.,  $x(n) = \frac{1}{2\pi} \int_{2\pi} X(\Omega)e^{j\Omega n} d\Omega$ ) will always converge, since the integration interval is finite.

## DTFT Examples and Exercises

**Example:** Find the DTFT of the signal  $x[n] = \alpha^n u[n]$ . Sketch the magnitude and phase spectrum assuming  $\alpha$  to be real-valued.

*Solution:* The DTFT is given by

$$\begin{aligned} X(e^{j\Omega}) &= \sum_{n=-\infty}^{\infty} \alpha^n u[n] e^{-j\Omega n} = \sum_{n=0}^{\infty} \alpha^n e^{-j\Omega n} \\ &= \frac{1}{1 - \alpha e^{-j\Omega}}, \quad |\alpha| < 1. \end{aligned}$$

If  $\alpha$  is real-valued and  $|\alpha| < 1$ , then we have

$$X(e^{j\Omega}) = \frac{1}{1 - \alpha \cos(\Omega) + j\alpha \sin(\Omega)}$$

## DTFT Examples and Exercises

$$\Rightarrow |X(e^{j\Omega})| = \frac{1}{\left((1 - \alpha \cos(\Omega))^2 + \alpha^2 \sin^2(\Omega)\right)^{1/2}}, \quad \text{and}$$

$$\arg\{X(e^{j\Omega})\} = -\tan^{-1}\left(\frac{\alpha \sin(\Omega)}{1 - \alpha \cos(\Omega)}\right)$$

*Can you draw the magnitude and phase plot? (see next slide)*

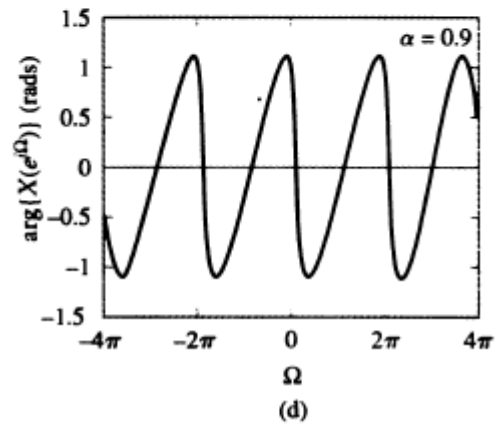
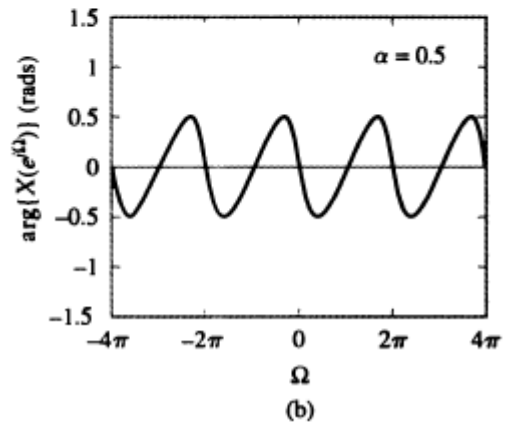
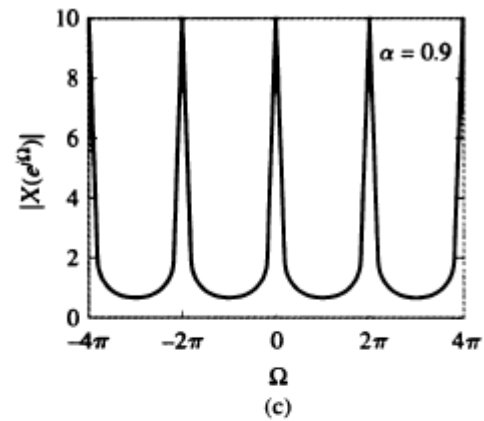
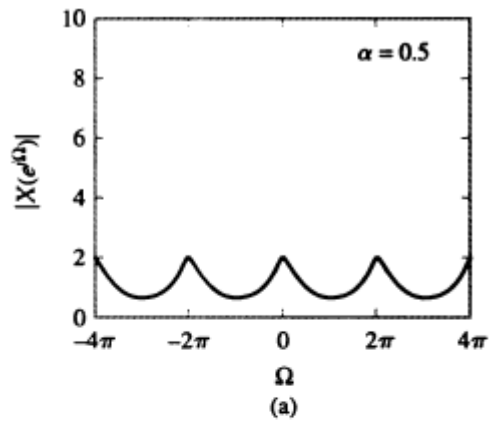
**Exercise:** Determine the DTFT of the following signals:

$$(i) x[n] = \delta[n], \quad (ii) x[n] = 1, \quad (iii) \cos(\Omega_0 n), \quad (iv) \sin(\Omega_0 n).$$

**Exercise:** Determine the inverse DTFT of the spectrum

$$X(e^{j\Omega}) = \begin{cases} 1, & |\Omega| < W \\ 0, & W < |\Omega| < \pi \end{cases}.$$

## Periodic Magnitude and Phase Spectra



## DTFT Properties

## Periodicity

- Recall the definition of the Fourier transform  $X$  of the sequence  $x$ :

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

- For all integer  $k$ , we have that

$$\begin{aligned} X(\Omega + 2\pi k) &= \sum_{n=-\infty}^{\infty} x(n)e^{-j(\Omega + 2\pi k)n} \\ &= \sum_{n=-\infty}^{\infty} x(n)e^{-j(\Omega n + 2\pi kn)} \\ &= \sum_{n=-\infty}^{\infty} x(n)e^{-j\Omega n} \\ &= X(\Omega). \end{aligned}$$

- Thus, the Fourier transform  $X$  of the sequence  $x$  is always *2π-periodic*.

## Linearity

- If  $x_1(n) \xleftrightarrow{\text{DTFT}} X_1(\Omega)$  and  $x_2(n) \xleftrightarrow{\text{DTFT}} X_2(\Omega)$ , then

$$a_1x_1(n) + a_2x_2(n) \xleftrightarrow{\text{DTFT}} a_1X_1(\Omega) + a_2X_2(\Omega),$$

where  $a_1$  and  $a_2$  are arbitrary complex constants.

- This is known as the **linearity property** of the Fourier transform.



## Translation or Time Shifting

- If  $x(n) \xleftrightarrow{\text{DTFT}} X(\Omega)$ , then

$$x(n - n_0) \xleftrightarrow{\text{DTFT}} e^{-j\Omega n_0} X(\Omega),$$

where  $n_0$  is an arbitrary integer.

- This is known as the **translation (or time-domain shifting) property** of the Fourier transform.

## Modulation or Frequency Shifting

- If  $x(n) \xleftrightarrow{\text{DTFT}} X(\Omega)$ , then

$$e^{j\Omega_0 n} x(n) \xleftrightarrow{\text{DTFT}} X(\Omega - \Omega_0),$$

where  $\Omega_0$  is an arbitrary real constant.

- This is known as the **modulation (or frequency-domain shifting) property** of the Fourier transform.

## Conjugation and Time Reversal

- If  $x(n) \xleftrightarrow{\text{DTFT}} X(\Omega)$ , then

$$x^*(n) \xleftrightarrow{\text{DTFT}} X^*(-\Omega).$$

- This is known as the **conjugation property** of the Fourier transform.
- If  $x(n) \xleftrightarrow{\text{DTFT}} X(\Omega)$ , then

$$x(-n) \xleftrightarrow{\text{DTFT}} X(-\Omega).$$

- This is known as the **time-reversal property** of the Fourier transform.

## Convolution

- If  $x_1(n) \xleftrightarrow{\text{DTFT}} X_1(\Omega)$  and  $x_2(n) \xleftrightarrow{\text{DTFT}} X_2(\Omega)$ , then

$$x_1 * x_2(n) \xleftrightarrow{\text{DTFT}} X_1(\Omega)X_2(\Omega).$$

- This is known as the **convolution (or time-domain convolution) property** of the Fourier transform.
- In other words, a convolution in the time domain becomes a multiplication in the frequency domain.
- This suggests that the Fourier transform can be used to avoid having to deal with convolution operations.

## Parseval's Relation

- If  $x(n) \xleftrightarrow{\text{DTFT}} X(\Omega)$ , then

$$\sum_{n=-\infty}^{\infty} |x(n)|^2 = \frac{1}{2\pi} \int_{2\pi} |X(\Omega)|^2 d\Omega$$

(i.e., the energy of  $x$  and energy of  $X$  are equal up to a factor of  $2\pi$ ).

- This is known as **Parseval's relation**.
- Since energy is often a quantity of great significance in engineering applications, it is extremely helpful to know that the Fourier transform *preserves energy* (up to a scale factor).

## Other Properties

- For a sequence  $x$  with Fourier transform  $X$ , the following assertions hold:
  - 1  $x$  is even  $\Leftrightarrow X$  is even; and
  - 2  $x$  is odd  $\Leftrightarrow X$  is odd.
- In other words, the forward and inverse Fourier transforms preserve even/odd symmetry.

## Other Properties

- A sequence  $x$  is *real* if and only if its Fourier transform  $X$  satisfies

$$X(\Omega) = X^*(-\Omega) \text{ for all } \Omega$$

(i.e.,  $X$  is *conjugate symmetric*).

- Thus, for a real-valued sequence, the portion of the graph of a Fourier transform for negative values of frequency  $\Omega$  is *redundant*, as it is completely determined by symmetry.
- From properties of complex numbers, one can show that  $X(\Omega) = X^*(-\Omega)$  is equivalent to

$$|X(\Omega)| = |X(-\Omega)| \quad \text{and} \quad \arg X(\Omega) = -\arg X(-\Omega)$$

(i.e.,  $|X(\Omega)|$  is *even* and  $\arg X(\Omega)$  is *odd*).

- Note that  $x$  being real does *not* necessarily imply that  $X$  is real.

## DT LTI Systems Given By Difference Equations

- Many LTI systems of practical interest can be represented using an *Nth-order linear difference equation with constant coefficients*.
- Consider a system with input  $x$  and output  $y$  that is characterized by an equation of the form

$$\sum_{k=0}^N b_k y(n-k) = \sum_{k=0}^M a_k x(n-k).$$

- Let  $h$  denote the impulse response of the system, and let  $X$ ,  $Y$ , and  $H$  denote the Fourier transforms of  $x$ ,  $y$ , and  $h$ , respectively.
- One can show that  $H(\Omega)$  is given by

$$H(\Omega) = \frac{Y(\Omega)}{X(\Omega)} = \frac{\sum_{k=0}^M a_k (e^{j\Omega})^{-k}}{\sum_{k=0}^N b_k (e^{j\Omega})^{-k}} = \frac{\sum_{k=0}^M a_k e^{-jk\Omega}}{\sum_{k=0}^N b_k e^{-jk\Omega}}.$$

- Each of the numerator and denominator of  $H$  is a *polynomial* in  $e^{-j\Omega}$ .
- Thus,  $H$  is a *rational function* in the variable  $e^{-j\Omega}$ .



## DTFT Properties: Examples and Exercises

**Example:** Find the output of a DT LTI system whose impulse response is given by  $h[n] = (1/(\pi n)) \sin(\pi n/2)$  if the input is  $x[n] = (1/2)^n u[n]$ .

*Solution:* The DTFT of the input is given by

$$X(e^{j\Omega}) = \frac{1}{1 - (1/2)e^{-j\Omega}}.$$

We have

$$\begin{aligned} \frac{\sin(\pi n)}{\pi n} &= \text{sinc}(n) \xleftrightarrow{\text{DTFT}} \text{rect}(\Omega/(2\pi)) \\ \Rightarrow h[n] &= \frac{\sin(\pi n/2)}{\pi n} = (1/2)\text{sinc}(n/2) \xleftrightarrow{\text{DTFT}} \text{rect}(\Omega/\pi) = H(e^{j\Omega}) \end{aligned}$$

## DTFT Properties: Examples and Exercises

Therefore, the DTFT of the output is given by

$$\begin{aligned} Y(e^{j\Omega}) &= X(e^{j\Omega})H(e^{j\Omega}) = \frac{\text{rect}(\Omega/\pi)}{1 - (1/2)e^{-j\Omega}} \\ &= \begin{cases} \frac{1}{1 - (1/2)e^{-j\Omega}} & |\Omega| \leq \pi/2 \\ 0, & \pi/2 < |\Omega| \leq \pi \end{cases} \end{aligned}$$

**Example:** Consider the two-path communication channel given by the input-output relationship

$$y[n] = x[n] + ax[n - 1], \quad |a| < 1.$$

Find the impulse response of the inverse system by the convolution property of DTFT.

## DTFT Properties: Examples and Exercises

*Solution: The impulse response of the system is*

$$h[n] = \delta[n] + a\delta[n - 1].$$

*The frequency response of the system is*

$$H(e^{j\Omega}) = 1 + ae^{-j\Omega},$$

*which implies that*

$$H^{inv}(e^{j\Omega}) = \frac{1}{1 + ae^{-j\Omega}} \quad \Rightarrow \quad h^{inv}[n] = (-a)^n u[n].$$

## DTFT Properties: Examples and Exercises

**Example:** Use the frequency-differentiation property and find out the DTFT of the signal

$$x[n] = (n + 1)\alpha^n u[n], \quad |\alpha| < 1.$$

**Solution:** We have

$$\alpha^n u[n] \xleftrightarrow{\text{DTFT}} \frac{1}{1 - \alpha e^{-j\Omega}}$$

$$\Rightarrow n\alpha^n u[n] \xleftrightarrow{\text{DTFT}} j \frac{d}{d\Omega} \left( \frac{1}{1 - \alpha e^{-j\Omega}} \right) = \frac{\alpha e^{-j\Omega}}{(1 - \alpha e^{-j\Omega})^2}$$

$$\Rightarrow (n + 1)\alpha^n u[n] \xleftrightarrow{\text{DTFT}} \frac{1}{1 - \alpha e^{-j\Omega}} + \frac{\alpha e^{-j\Omega}}{(1 - \alpha e^{-j\Omega})^2} = \frac{1}{(1 - \alpha e^{-j\Omega})^2}$$

## DTFT Table of Properties

Property	Time Domain	Frequency Domain
Linearity	$a_1x_1(n) + a_2x_2(n)$	$a_1X_1(\Omega) + a_2X_2(\Omega)$
Translation	$x(n - n_0)$	$e^{-j\Omega n_0}X(\Omega)$
Modulation	$e^{j\Omega_0 n}x(n)$	$X(\Omega - \Omega_0)$
Conjugation	$x^*(n)$	$X^*(-\Omega)$
Time Reversal	$x(-n)$	$X(-\Omega)$
Upsampling	$(\uparrow M)x(n)$	$X(M\Omega)$
Downsampling	$(\downarrow M)x(n)$	$\frac{1}{M} \sum_{k=0}^{M-1} X\left(\frac{\Omega - 2\pi k}{M}\right)$
Convolution	$x_1 * x_2(n)$	$X_1(\Omega)X_2(\Omega)$
Multiplication	$x_1(n)x_2(n)$	$\frac{1}{2\pi} \int_{2\pi} X_1(\theta)X_2(\Omega - \theta)d\theta$
Freq.-Domain Diff.	$nx(n)$	$j\frac{d}{d\Omega}X(\Omega)$
Differencing	$x(n) - x(n - 1)$	$(1 - e^{-j\Omega})X(\Omega)$
Accumulation	$\sum_{k=-\infty}^n x(k)$	$\frac{e^{j\Omega}}{e^{j\Omega} - 1}X(\Omega) + \pi X(0) \sum_{k=-\infty}^{\infty} \delta(\Omega - 2\pi k)$

## DTFT Table of Properties

Property	
Periodicity	$X(\Omega) = X(\Omega + 2\pi)$
Parseval's Relation	$\sum_{n=-\infty}^{\infty}  x(n) ^2 = \frac{1}{2\pi} \int_{2\pi}  X(\Omega) ^2 d\Omega$
Even Symmetry	$x \text{ is even} \Leftrightarrow X \text{ is even}$
Odd Symmetry	$x \text{ is odd} \Leftrightarrow X \text{ is odd}$
Real / Conjugate Symmetry	$x \text{ is real} \Leftrightarrow X \text{ is conjugate symmetric}$

## DTFT of Common DT Signals

Pair	$x(n)$	$X(\Omega)$
1	$\delta(n)$	1
2	1	$2\pi \sum_{k=-\infty}^{\infty} \delta(\Omega - 2\pi k)$
3	$u(n)$	$\frac{e^{j\Omega}}{e^{j\Omega} - 1} + \sum_{k=-\infty}^{\infty} \pi \delta(\Omega - 2\pi k)$
4	$a^n u(n),  a  < 1$	$\frac{e^{j\Omega}}{e^{j\Omega} - a}$
5	$-a^n u(-n-1),  a  > 1$	$\frac{e^{j\Omega}}{e^{j\Omega} - a}$
6	$a^{ n },  a  < 1$	$\frac{1-a^2}{1-2a\cos\Omega+a^2}$
7	$\cos \Omega_0 n$	$\pi \sum_{k=-\infty}^{\infty} [\delta(\Omega - \Omega_0 - 2\pi k) + \delta(\Omega + \Omega_0 - 2\pi k)]$
8	$\sin \Omega_0 n$	$j\pi \sum_{k=-\infty}^{\infty} [\delta(\Omega + \Omega_0 - 2\pi k) - \delta(\Omega - \Omega_0 - 2\pi k)]$
9	$(\cos \Omega_0 n)u(n)$	$\frac{e^{j2\Omega} - e^{j\Omega \cos \Omega_0}}{e^{j2\Omega} - 2e^{j\Omega \cos \Omega_0} + 1} + \frac{\pi}{2} \sum_{k=-\infty}^{\infty} [\delta(\Omega - 2\pi k - \Omega_0) + \delta(\Omega - 2\pi k + \Omega_0)]$
10	$(\sin \Omega_0 n)u(n)$	$\frac{e^{j\Omega} \sin \Omega_0}{e^{j2\Omega} - 2e^{j\Omega \cos \Omega_0} + 1} + \frac{\pi}{2j} \sum_{k=-\infty}^{\infty} [\delta(\Omega - 2\pi k - \Omega_0) - \delta(\Omega - 2\pi k + \Omega_0)]$
11	$\frac{B}{\pi} \text{sinc } Bn, 0 < B < \pi$	$\sum_{k=-\infty}^{\infty} \text{rect} \left( \frac{\Omega - 2\pi k}{2B} \right)$
12	$u(n) - u(n-M)$	$e^{-j\Omega(M-1)/2} \left( \frac{\sin(M\Omega/2)}{\sin(\Omega/2)} \right)$

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