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

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• A sequence X is real if and only if its Fourier transform X satisfies

 $X(\Omega) = X^*(-\Omega)$ for all Ω

)i.e., X has conjugate symmetry.(

- Thus, for a real-valued sequence, the portion of the graph of a Fourier transform for negative values of frequency Ω 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

 $|X(\Omega)| = |X(-\Omega)|$ and $\arg X(\Omega) = -\arg X(-\Omega)$

)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.

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• The DTFT analysis and synthesis equations are, respectively, given by

$$X(\Omega = \left(\sum_{k^{\infty}=1}^{\infty} x(k) e^{-jk\Omega} \text{ and } x(P = \left(\frac{1}{2\pi} \sum_{2\pi}^{N} X(\Omega) e^{jn\Omega} d\Omega\right)\right)$$

• The CTFS synthesis and analysis equations are, respectively, given by

$$x_{\rm c}(t=(\sum_{k^{\infty}=1}^{\infty}a(k)e^{jk(2\pi/T)t}) \text{ and } a(n)=\frac{1}{7}x_{\rm c}(t)e^{-jn(2\pi/T)t}dt$$

which can be rewritten, respectively, as

$$x_{c}(t = (\sum_{k^{\infty}=1}^{\infty} a(-k)e^{-jk(2\pi/T)t}) \text{ and } a(-n) = \frac{1}{T} x_{c}(t)e^{jn(2\pi/T)t}dt.$$

- The CTFS synthesis equation with $T = 2\pi$ corresponds to the DTFT analysis equation with $X = x_c$, $\Omega = t$, and x(n) = a(-n.)
- The CTFS analysis equation with $T = 2\pi$ corresponds to the DTFT synthesis equation with $X = x_c$ and x(n) = a(-n.)
- Consequently, the DTFT X of the sequence x can be viewed as a CTFS representation of the 2 π -periodic spectrum X.

- The Fourier transform can be generalized to also handle periodic signals.
- Consider an N-periodic sequence X.
- Define the sequence X_N as

$$x_{N}(n = (\begin{array}{c} x(n) & \text{for } 0 \leq n < N \\ 0 & \text{otherwise.} \end{array}$$

) i.e., $X_N(n)$ is equal to X(n) over a single period and zero elsewhere).

- Let a denote the Fourier series coefficient sequence of X.
- Let X and X_W denote the Fourier transforms of X and X_W , respectively.
- The following relationships can be shown to hold:

$$X(\Omega) = \frac{2\pi}{N} \sum_{k^{\infty} - =}^{\infty} X_N \frac{2\pi k}{N} \delta \Omega - \frac{2\pi k}{N} ,$$
$$a_k = \frac{1}{N} X_N \frac{2\pi k}{N} \quad \text{and} \quad X(\Omega) = 2\pi \sum_{k^{\infty} - =}^{\infty} a_k \delta \Omega - \frac{2\pi k}{N} .$$

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- The Fourier series coefficient sequence a is produced by sampling X_N at integer multiples of the fundamental frequency $2\pi \frac{1}{N}$ nd scaling the resulting sequence by $\frac{1}{N}$
- The Fourier transform of a periodic sequence can only be nonzero at integer multiples of the fundamental frequency.

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Section 10.4

Fourier Transform and Frequency Spectra of Signals

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- Like Fourier series, the Fourier transform also provides us with a frequency-domain perspective on sequences.
- That is, instead of viewing a sequence as having information distributed with respect to *time*(i.e., a function whose domain is time), we view a sequence as having information distributed with respect to *frequency*(i.e., a function whose domain is frequency).
- The Fourier transform X of a sequence X provides a means to *quantify* how much information X has at different frequencies.
- The distribution of information in a sequence over different frequencies is referred to as the *frequency spectrum* of the sequence.

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• To gain further insight into the role played by the Fourier transform X in the context of the frequency spectrum of x, it is helpful to write the Fourier transform representation of x with $X(\Omega)$ expressed in *polar form* as follows:

$$X(n) = \frac{1}{2\pi} \left\{ \begin{array}{c} & & \\ & \chi(\Omega) \, e^{i\Omega n} d\Omega = \frac{1}{2\pi} \\ & & 2\pi \end{array} \right\} |X(\Omega)| \, e^{i[\Omega n + \arg X(\Omega)]} d\Omega.$$

- In effect, the quantity $|X(\Omega)|$ is a *weight* that determines how much the complex sinusoid at frequency Ω contributes to the integration result X(n)
- Perhaps, this can be more easily seen if we express the above integral as the *limit of a sum*, derived from an approximation of the integral using the area of rectangles, as shown on the next slide. [Recall that $\int_{a}^{b} f(x) dx = \lim_{n \to \infty} \sum_{k=1}^{n} f(x_k) \Delta x$ where $\Delta x = \frac{b-a}{n}$ and $x_k = a + k\Delta x$ [.

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• Expressing the integral (from the previous slide) as the *limit of a sum*, we obtain

$$x(n) = \lim_{\ell \to \infty} \frac{1}{2\pi} \sum_{k=1}^{\ell} \Delta \Omega \, \frac{1}{X}(\Omega') \, \frac{1}{\ell} e^{I[\Omega' n + \arg X(\Omega')]},$$

where $\Delta \Omega = \frac{2\pi}{l}$ and $\Omega' = k \Delta \Omega$.

- In the above equation, the *k*th term in the summation corresponds to a complex sinusoid with fundamental frequency $\Omega' = k\Delta\Omega$ that has had its *amplitude scaled* by a factor of $|X(\Omega')|$ and has been *time shifted* by an amount that depends on $\arg X(\Omega')$.
- For a given $\Omega' = k\Delta\Omega$ (which is associated with the *k*th term in the summation), the larger $|X(\Omega')|$ is, the larger the amplitude of its corresponding complex sinusoid $\Theta^{\Omega' n}$ will be, and therefore the larger the contribution the *k*th term will make to the overall summation.
- In this way, we can use $|X(\Omega')|$ as a *measure* of how much information a sequence X has at the frequency Ω .

- The Fourier transform X of the sequence X is referred to as the frequency spectrum of X.
- The magnitude $|X(\Omega)|$ of the Fourier transform X is referred to as the magnitude spectrum of X.
- The argument $\arg X(\Omega)$ of the Fourier transform X is referred to as the phase spectrum of X.
- Since the Fourier transform is a function of a real variable, a sequence can potentially have information at any real frequency.
- Earlier, we saw that for periodic sequences, the Fourier transform can only be nonzero at integer multiples of the fundamental frequency.
- So, the Fourier transform and Fourier series give a consistent picture in terms of frequency spectra.
- Since the frequency spectrum is complex (in the general case), it is *usually represented using two plots*, one showing the magnitude spectrum and one showing the phase spectrum.

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• Recall that, for a *real* sequence *X*, the Fourier transform *X* of *X* satisfies $X(\Omega) = X^*(-\Omega)$

(i.e., X is conjugate symmetric), which is equivalent to

 $|X(\Omega)| = |X(-\Omega)|$ and $\arg X(\Omega) = -\arg X(-\Omega)$.

- Since $|X(\Omega)| = |X(-\Omega)|$, the magnitude spectrum of a *real* sequence is always *even*.
- Similarly, since $\arg X(\Omega) = -\arg X(-\Omega)$, the phase spectrum of a *real* sequence is always *odd*.
- Due to the symmetry in the frequency spectra of real sequences, we typically *ignore negative frequencies* when dealing with such sequences.
- In the case of sequences that are complex but not real, frequency spectra do not possess the above symmetry, and *negative frequencies become important*.

- A sequence X with Fourier transform X is said to be bandlimited if, for some nonnegative real constant $B, X(\Omega) = 0$ for all Ω satisfying $|\Omega| > B$.
- In the context of real sequences, we usually refer to B as the bandwidth of the signal X.
- The (real) sequence with the Fourier transform X shown below has bandwidth B.



One can show that a sequence *cannot be both time limited and bandlimited*. (This follows from the time/frequency scaling property of the Fourier transform.)

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Section 10.5

Fourier Transform and LTISystems

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- Consider a ITI system with input X, output Y, and impulse response h, and let X, Y, and H denote the Fourier transforms of X, Y, and h, respectively.
- Since y(n) = x * h(n), we have that

$$Y(\Omega) = X(\Omega) H(\Omega).$$

- The function H is called the frequency response of the system.
- A ITI system is *completely characterized* by its frequency response *H*. The
- above equation provides an alternative way of viewing the behavior of a ITI system. That is, we can view the system as operating in the frequency domain on the Fourier transforms of the input and output signals.
- The frequency spectrum of the output is the product of the frequency spectrum of the input and the frequency response of the system.

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- In the general case, the frequency response *H* is a complex-valued function.
- Often, we represent $H(\Omega)$ in terms of its magnitude $|H(\Omega)|$ and argument arg $H(\Omega)$.
- The quantity $|H(\Omega)|$ is called the magnitude response of the system.
- The quantity $\arg H(\Omega)$ is called the phase response of the system.
- Since $Y(\Omega) = X(\Omega) H(\Omega)$, we trivially have that

 $|Y(\Omega)| = |X(\Omega)| |H(\Omega)|$ and $\arg Y(\Omega) = \arg X(\Omega) + \arg H(\Omega)$.

- The magnitude spectrum of the output equals the magnitude spectrum of the input times the magnitude response of the system.
- The phase spectrum of the output equals the phase spectrum of the input plus the phase response of the system.

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