


Lecture 4

Frequency Domain Representation of Discrete Time Signal

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- The frequency domain representation of discrete time sequence is the **discrete-time Fourier transform (DTFT)**. DTFT is a frequency analysis tool for aperiodic discrete-time signals
 - This transform maps a time-domain sequence into a continuous function of the frequency variable ω .

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Review of Continuous-Time Fourier Transform (CTFT)

- CTFT:

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

Fourier Spectrum
Or Spectrum

- Inverse CTFT:

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

$$x_a(t) \xleftrightarrow{\text{CTFT}} X_a(j\Omega)$$

Definition of Discrete-Time Fourier Transform (DTFT)

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

DTFT

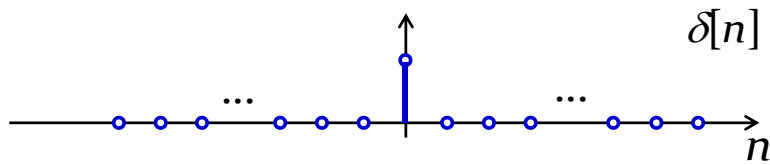
where, ω is a continuous variable in the range $-\infty < \omega < \infty$

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

Inverse DTFT

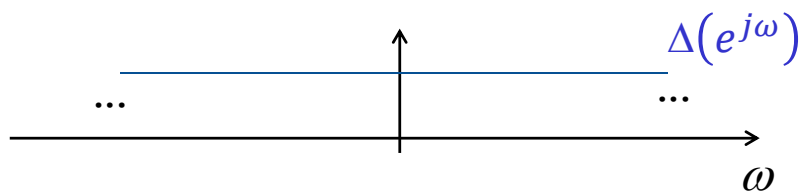
Why one is sum and the other integral?

Example 1



DTFT:

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



Example 2

- Causal sequence $x[n] = \alpha^n \mu[n]$, $|\alpha| < 1$

Its DTFT is given by

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

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

Recall: $1 + q + q^2 + \dots + q^\infty = \frac{1}{1 - q}$ for $|q| < 1$

DTFT

- In general, $X(e^{j\omega})$ is a complex function of the real variable ω , and can be written as

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

- $X(e^{j\omega})$ can alternatively be expressed as

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

where, $\theta(\omega) = \arg\{X(e^{j\omega})\}$

- $|X(e^{j\omega})|$ is called the magnitude function
- $\theta(\omega)$ is called the phase function
- In applications where DTFT is called Fourier spectrum, $|X(e^{j\omega})|$ and $\theta(\omega)$ are called magnitude and phase spectra

Symmetry of DTFT

- For a **real sequence** $x[n]$, $|X(e^{j\omega})|$ and $X_{\text{re}}(e^{j\omega})$ are even functions of ω , whereas $\theta(\omega)$ and $X_{\text{im}}(e^{j\omega})$ are odd functions of ω .

- Proof:
$$\begin{aligned} X(e^{j\omega}) &= \sum_{n=-\infty}^{\infty} x[n]e^{-j\omega n} \\ X(e^{-j\omega}) &= \sum_{n=-\infty}^{\infty} x[n]e^{j\omega n} \\ &= \left\{ \sum_{n=-\infty}^{\infty} x[n]e^{-j\omega n} \right\}^* , \text{ for real } x[n] \\ &= X^*(e^{j\omega}) \end{aligned}$$

- Therefore, $|X(e^{j\omega})| = |X(e^{-j\omega})|$ and $\theta(\omega) = -\theta(-\omega)$

Periodicity of DTFT

- $X(e^{j\omega}) = X(e^{j(\omega+2k\pi)})$, i. e. $|X(e^{j\omega})|e^{j[\theta(\omega)+2k\pi]} = |X(e^{j\omega})|e^{j\theta(\omega)}$ for any integer k .

- Proof:

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

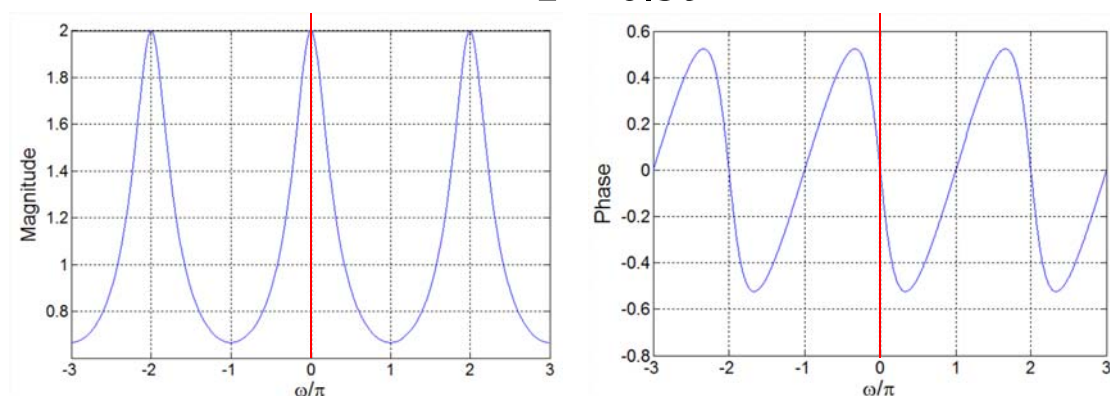
- In other words, the phase function $\theta(\omega)$ cannot be uniquely specified for any DTFT.
- Unless otherwise stated, we assume that the phase function $\theta(\omega)$ is restricted to the range of $-\pi < \theta(\omega) < \pi$

called the **principle value**.

Example 2 (Cont'd)

- The magnitude and phase of DTFT of

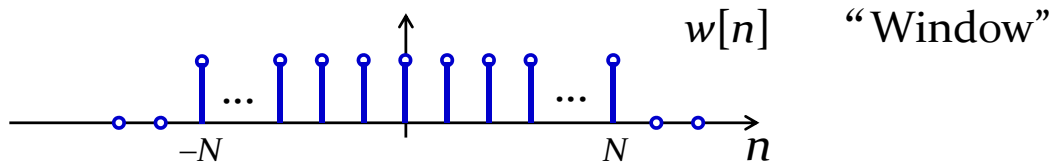
$$X(e^{j\omega}) = \frac{1}{1 - 0.5e^{-j\omega}}$$



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

$$\theta(\omega) = -\theta(-\omega)$$

Example 3



DTFT:

$$\begin{aligned} W(e^{j\omega}) &= \sum_{k=-\infty}^{\infty} w[k]e^{-j\omega k} = \sum_{k=-N}^N e^{-j\omega k} \\ &= e^{-j\omega N} (1 + e^{j\omega} + e^{j2\omega} + \dots + e^{j2N\omega}) \end{aligned}$$

Recall: $1 + q + q^2 + \dots + q^M = \frac{1 - q^{M+1}}{1 - q} \quad \begin{matrix} q = e^{j\omega} \\ M = 2N \end{matrix}$

Example 3 Cont.

DTFT: $W(e^{j\omega}) = e^{-j\omega N} (1 + e^{j\omega} + e^{j2\omega} + \dots + e^{j2N\omega})$

$$= e^{-j\omega N} \frac{1 - e^{j\omega(2N+1)}}{1 - e^{j\omega}}$$

$$= \frac{e^{-j\omega N} - e^{j\omega N} e^{j\omega}}{1 - e^{j\omega}}$$

$$= \frac{e^{-j\omega(N+\frac{1}{2})} - e^{j\omega(N+\frac{1}{2})}}{e^{-j\frac{\omega}{2}} - e^{j\frac{\omega}{2}}}$$

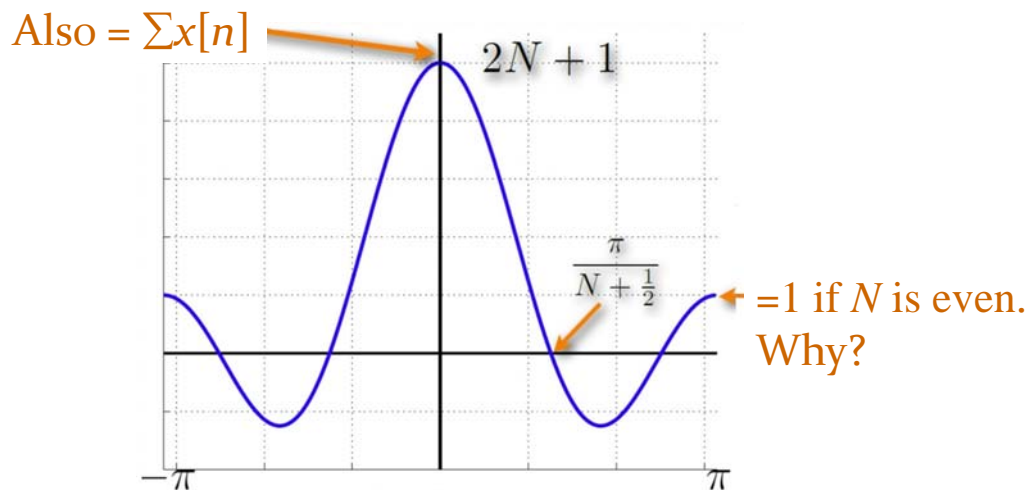
$$= \frac{\sin\left(\omega\left(N+\frac{1}{2}\right)\right)}{\sin\left(\frac{\omega}{2}\right)}$$

$$\times \frac{e^{-j\frac{\omega}{2}}}{e^{-j\frac{\omega}{2}}}$$

Periodic Sinc

Example 3 Cont.

$$W(e^{j\omega}) = \frac{\sin\left(\omega\left(N + \frac{1}{2}\right)\right)}{\sin\left(\frac{\omega}{2}\right)} \quad \rightarrow 2N+1 \text{ as } \omega \rightarrow 0$$



Example of Inverse DTFT

- Find the inverse DTFT of

$$H_{LP}(e^{j\omega}) = \begin{cases} 1, & 0 \leq |\omega| \leq \omega_c \\ 0, & \omega_c \leq |\omega| \leq \pi \end{cases}$$

- A:

$$\begin{aligned} h_{LP}[n] &= \frac{1}{2\pi} \int_{-\pi}^{\pi} H_{LP}(e^{j\omega}) e^{j\omega n} d\omega = \frac{1}{2\pi} \int_{-\omega_c}^{\omega_c} e^{j\omega n} d\omega \\ &= \frac{1}{2\pi} \left(\frac{e^{j\omega_c n}}{jn} - \frac{e^{-j\omega_c n}}{jn} \right) = \frac{\sin \omega_c n}{\pi n}, \text{ for } n \neq 0 \\ h_{LP}[0] &= \frac{1}{2\pi} \int_{-\pi}^{\pi} H_{LP}(e^{j\omega}) d\omega = \frac{1}{2\pi} \int_{-\omega_c}^{\omega_c} d\omega = \frac{\omega_c}{\pi} \end{aligned}$$

Properties of the DTFT

- **Linearity:**

Let $g[n] \leftrightarrow G(e^{j\omega})$ and $h[n] \leftrightarrow H(e^{j\omega})$

Then $\alpha g[n] + \beta h[n] \leftrightarrow \alpha G(e^{j\omega}) + \beta H(e^{j\omega})$

- **Periodicity:** $X(e^{j(\omega+2\pi)}) = X(e^{j\omega})$

Properties of the DTFT Cont.

- **Time Reversal:**

Let $x[n] \leftrightarrow X(e^{j\omega})$

Then $x[-n] \leftrightarrow X(e^{-j\omega})$

$= X^*(e^{j\omega})$ if $x[n]$ is real

If $x[n]=x[-n]$ and $x[n]$ is real, then

$$X(e^{j\omega}) = X^*(e^{j\omega}) \rightarrow X(e^{j\omega}) \text{ is real}$$

- Q: Suppose $x[n] \leftrightarrow X(e^{j\omega})$, $x[n] \in \mathcal{Real}$

$$? \leftrightarrow \mathcal{Re}\{X(e^{j\omega})\}$$

- A: Decompose $x[n]$ to even and odd functions

$$x[n] = x_e[n] + x_o[n],$$

where
$$x_e[n] = \frac{1}{2}(x[n] + x[-n])$$

$$x_o[n] = \frac{1}{2}(x[n] - x[-n])$$

$$\boxed{x_e[n]} \leftrightarrow X_e(e^{j\omega}) = \frac{1}{2}(X(e^{j\omega}) + X(e^{-j\omega}))$$

$$= \frac{1}{2}(X(e^{j\omega}) + X^*(e^{j\omega})) = \mathcal{Re}\{X(e^{j\omega})\}$$

$$x_o[n] \leftrightarrow X_o(e^{j\omega}) = \frac{1}{2}(X(e^{j\omega}) - X(e^{-j\omega})) = j\mathcal{Im}\{X(e^{j\omega})\}$$

Properties of the DTFT Cont.

- Time and Frequency Shifting

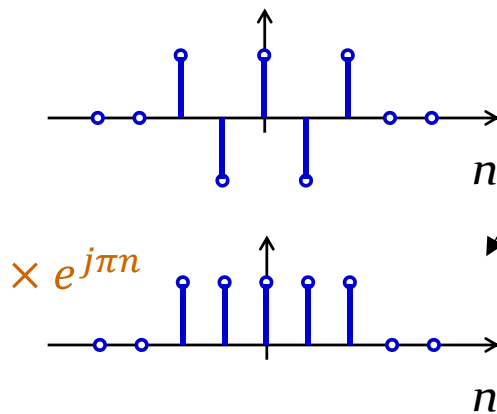
Let
$$x[n] \leftrightarrow X(e^{j\omega})$$

Then
$$x[n - n_d] \leftrightarrow e^{-j\omega n_d} X(e^{j\omega})$$

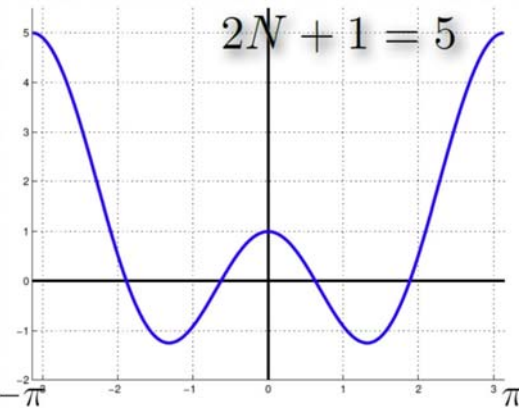
$$e^{j\omega_0 n} x[n] \leftrightarrow X(e^{j(\omega - \omega_0)})$$

Example 4

What is the DTFT of:



$$W(e^{j\omega}) = \frac{\sin\left(\omega\left(N + \frac{1}{2}\right)\right)}{\sin\left(\frac{\omega}{2}\right)}$$



$$W(e^{j\omega}) = \frac{\sin\left((\omega - \pi)\left(N + \frac{1}{2}\right)\right)}{\sin\left(\frac{\omega - \pi}{2}\right)}$$

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Properties of the DTFT Cont.

- Differentiation in frequency

Let $x[n] \leftrightarrow X(e^{j\omega})$

Then $nx[n] \leftrightarrow j \frac{dX(e^{j\omega})}{d\omega}$

Proof: $X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x[n]e^{-j\omega n}$

Differentiate both side to get $\frac{dX(e^{j\omega})}{d\omega} = \sum_{n=-\infty}^{\infty} -jnx[n]e^{-j\omega n}$

Multiply both side by j , we get $j \frac{dX(e^{j\omega})}{d\omega} = \sum_{n=-\infty}^{\infty} nx[n]e^{-j\omega n}$

Example 5

- Determine DTFT $Y(e^{j\omega})$ of
$$y[n] = (n + 1)\alpha^n \mu[n], \quad |\alpha| < 1$$

- Let $x[n] = \alpha^n \mu[n]$, $|\alpha| < 1$

- We can therefore write

$$y[n] = nx[n] + x[n]$$

- From example 2, we have known that the DTFT of $x[n]$ is given by

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

- Using the differentiation in frequency, we observe that DTFT of $nx[n]$ is given by,

$$j \frac{dX(e^{j\omega})}{d\omega} = j \frac{d}{d\omega} \left(\frac{1}{1 - \alpha e^{-j\omega}} \right) = \frac{\alpha e^{-j\omega}}{(1 - \alpha e^{-j\omega})^2}$$

- Next, using linear property of DTFT, we arrive at

$$Y(e^{j\omega}) = \frac{\alpha e^{-j\omega}}{(1 - \alpha e^{-j\omega})^2} + \frac{1}{1 - \alpha e^{-j\omega}} = \frac{1}{(1 - \alpha e^{-j\omega})^2}$$

Properties of the DTFT Cont.

- **Convolution**

Let $x[n] \leftrightarrow X(e^{j\omega})$ and $h[n] \leftrightarrow H(e^{j\omega})$

- DTFT convolution theorem

If $y[n] = x[n] \otimes h[n]$

Then $y[n] \leftrightarrow Y(e^{j\omega}) = X(e^{j\omega})H(e^{j\omega})$

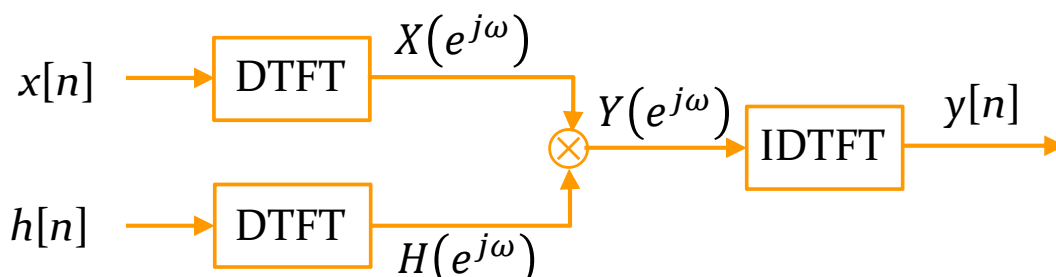
- **DTFT convolution modulation**

If $y[n] = x[n]h[n]$

Then $y[n] \leftrightarrow Y(e^{j\omega}) = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\theta})H(e^{j(\omega-\theta)})d\theta$

Linear Convolution Using DTFT

- Linear convolution $y[n]$ of the sequence $x[n]$ and $h[n]$ can be performed as follows:
 - Compute the DTFTs $X(e^{j\omega})$ and $H(e^{j\omega})$ of the sequences $x[n]$ and $h[n]$, respectively.
 - Form DTFT $Y(e^{j\omega}) = X(e^{j\omega})H(e^{j\omega})$
 - Compute the IDTFT $y[n]$ of $Y(e^{j\omega})$



Properties of the DTFT Cont.

- Parseval's theorem

Let $x[n] \leftrightarrow X(e^{j\omega})$ $h[n] \leftrightarrow H(e^{j\omega})$

Then
$$\sum_{n=-\infty}^{\infty} x[n]h^*[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\omega})H^*(e^{j\omega})d\omega$$

Proof:
$$\begin{aligned} \sum_{n=-\infty}^{\infty} x[n]h^*[n] &= \sum_{n=-\infty}^{\infty} x[n] \left(\frac{1}{2\pi} \int_{-\pi}^{\pi} H^*(e^{j\omega})e^{-j\omega n}d\omega \right) \\ &= \frac{1}{2\pi} \int_{-\pi}^{\pi} H^*(e^{j\omega}) \left(\sum_{n=-\infty}^{\infty} x[n]e^{-j\omega n} \right) d\omega \\ &= \frac{1}{2\pi} \int_{-\pi}^{\pi} H^*(e^{j\omega}) X(e^{j\omega}) d\omega \end{aligned}$$

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Energy & Energy Density Spectrum

- Energy: $E_g = \sum_{n=-\infty}^{\infty} |x[n]|^2$
- According to Parseval's theorem, when $h[n] = x[n]$,

$$E_g = \sum_{n=-\infty}^{\infty} |x[n]|^2 = \sum_{n=-\infty}^{\infty} x[n]x^*[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(e^{j\omega})|^2 d\omega$$

- Energy density spectrum:

$$S_{xx} = \int_{-\pi}^{\pi} |X(e^{j\omega})|^2 d\omega$$

Energy Density Spectrum

- Example – Compute the energy of the sequence

$$h_{LP}[n] = \frac{\sin(\omega_c n)}{\pi n}, \quad -\infty < n < \infty$$

- Here

$$\sum_{n=-\infty}^{\infty} |h_{LP}[n]|^2 = \frac{1}{2\pi} \int_{-\pi}^{\pi} |H_{LP}(e^{j\omega t})|^2 d\omega$$

where

$$H_{LP}(e^{j\omega t}) = \begin{cases} 1, & 0 \leq |\omega| \leq \omega_c \\ 0, & \omega_c \leq |\omega| \leq \pi \end{cases}$$

- Therefore:

$$\sum_{n=-\infty}^{\infty} |h_{LP}[n]|^2 = \frac{1}{2\pi} \int_{-\omega_c}^{\omega_c} d\omega = \frac{\omega_c}{\pi} < \infty$$

Symmetry Relations If $x[n]$ is real

Sequence	Discrete-Time Fourier Transform
$x[n]$	$X(e^{j\omega})$
$x_{\text{ev}}[n]$	$X_{\text{re}}(e^{j\omega})$
$x_{\text{od}}[n]$	$jX_{\text{im}}(e^{j\omega})$
Conjugate Symmetric	$X(e^{j\omega}) = X^*(e^{-j\omega})$ $X_{\text{re}}(e^{j\omega}) = X_{\text{re}}(e^{-j\omega})$ $X_{\text{im}}(e^{j\omega}) = -X_{\text{im}}(e^{-j\omega})$ $ X(e^{j\omega}) = X(e^{-j\omega}) $ $\arg\{X(e^{j\omega})\} = -\arg\{X(e^{-j\omega})\}$

Symmetry Relations If $x[n]$ is complex

Sequence	Discrete-Time Fourier Transform
$x[n]$	$X(e^{j\omega})$
$x[-n]$	$X(e^{-j\omega})$
$x^*[-n]$	$X^*(e^{j\omega})$
$x^*[n]$	$X^*(e^{-j\omega})$
$x_{\text{re}}[n]$	$X_{\text{cs}}(e^{j\omega}) = \frac{1}{2}\{X(e^{j\omega}) + X^*(e^{-j\omega})\}$
$jx_{\text{im}}[n]$	$X_{\text{ca}}(e^{j\omega}) = \frac{1}{2}\{X(e^{j\omega}) - X^*(e^{-j\omega})\}$
$x_{\text{cs}}[n]$	$X_{\text{re}}(e^{j\omega})$
$x_{\text{ca}}[n]$	$jX_{\text{im}}(e^{j\omega})$

Band-Limited DT Signals

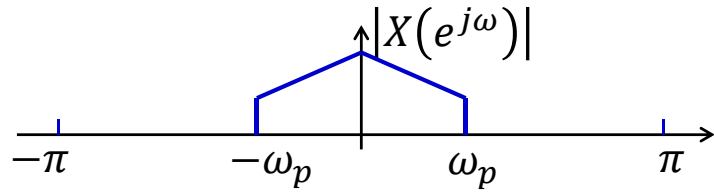
- The spectrum of a DT signal is a periodic function of ω with a period of 2π . Thus a full-band signal has a spectrum occupying the frequency range $-\pi < \omega \leq \pi$.
- A band-limited DT signal has a spectrum that is limited to a portion of the frequency range $-\pi < \omega \leq \pi$.
- An ideal band-limited signal:

$$X(e^{j\omega}) = \begin{cases} 0, & 0 \leq \omega < \omega_a \\ \text{non zero}, & \omega_a \leq \omega \leq \omega_b \\ 0, & \omega_b \leq \omega \leq \pi \end{cases}$$

Classification of Band-Limited Signal

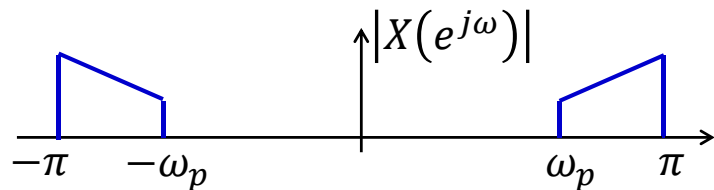
- Lowpass real signal

Bandwidth: ω_p



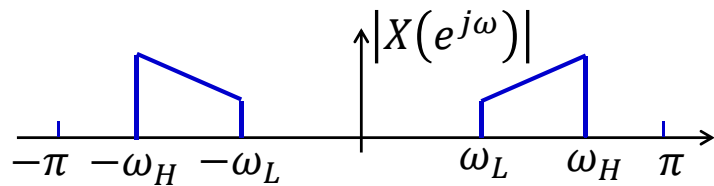
- Highpass real signal

Bandwidth: ω_p



- Bandpass real signal

Bandwidth: $\omega_H - \omega_L$



DTFT Convergence Condition

- The infinite series

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

may or may not converge.

- If it converges for all value of ω , then the DTFT $X(e^{j\omega})$ exists.
- Consider the finite sum:

$$X_K(e^{j\omega}) = \sum_{n=-K}^K x[n]e^{-j\omega n}$$

- Strong convergence: $X(e^{j\omega})$ converge uniformly, i.e.,

$$\lim_{K \rightarrow \infty} X_K(e^{j\omega}) = X(e^{j\omega})$$

- $x[n]$ absolutely summable $\Rightarrow X(e^{j\omega})$ exist and converge uniformly

$$\begin{aligned} \sum_{n=-\infty}^{\infty} |x[n]| < \infty &\Rightarrow |X(e^{j\omega})| = \left| \sum_{n=-\infty}^{\infty} x[n]e^{-j\omega n} \right| \\ &\leq \sum_{n=-\infty}^{\infty} |x[n]| |e^{-j\omega n}| = \sum_{n=-\infty}^{\infty} |x[n]| < \infty \end{aligned}$$

- This is a **sufficient condition**, not necessary.

Example

$x[n] = \alpha^n \mu[n]$, $|\alpha| < 1$ is absolutely summable, as

$$\sum_{n=-\infty}^{\infty} |\alpha^n \mu[n]| = \sum_{n=0}^{\infty} |\alpha^n| = \frac{1}{1 - |\alpha|} < \infty$$

and therefore is DTFT $X(e^{j\omega})$ converge to $\frac{1}{1 - \alpha e^{-j\omega}}$ uniformly.

Since
$$\sum_{n=-\infty}^{\infty} |x[n]|^2 \leq \left(\sum_{n=-\infty}^{\infty} |x[n]| \right)^2$$

- an absolutely summable sequence has always a finite energy,
- However, a finite energy sequence is not necessary absolutely summable.

Example: a sequence $x[n] = \frac{1}{n} \mu[n-1]$ has a finite energy, as $\sum_{n=1}^{\infty} \left(\frac{1}{n}\right)^2 = \frac{\pi^2}{6}$, however, it is not absolutely summable as $\sum_{n=1}^{\infty} \left|\frac{1}{n}\right| = \sum_{n=1}^{\infty} \frac{1}{n}$ does not converge.

- Weak convergence: $X(e^{j\omega})$ converge mean square

$$\lim_{K \rightarrow \infty} \int_{-\pi}^{\pi} |X(e^{j\omega}) - X_K(e^{j\omega})|^2 d\omega = 0$$

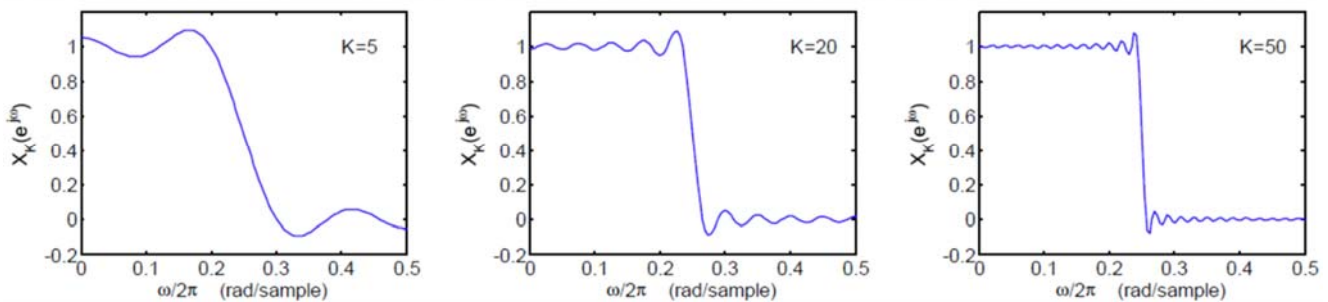
- $x[n]$ finite energy $\Rightarrow X(e^{j\omega})$ converge mean square.

$$\sum_{n=-\infty}^{\infty} |x[n]|^2 < \infty \Rightarrow \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(e^{j\omega})|^2 d\omega < \infty$$

- The absolute value of the error $|X(e^{j\omega}) - X_K(e^{j\omega})|$ may not go to zero when K goes to infinite.

Example

- $h_{LP}[n] = \frac{\sin 0.5\pi n}{\pi n} \leftrightarrow H_{LP}(e^{j\omega}) = \begin{cases} 1, & 0 \leq |\omega| \leq 0.5\pi \\ 0, & 0.5\pi \leq |\omega| \leq \pi \end{cases}$
- $\sum_{n=-\infty}^{\infty} |h_{LP}[n]|^2 = \frac{\omega_c}{\pi} < \infty$, but not absolutely summable.

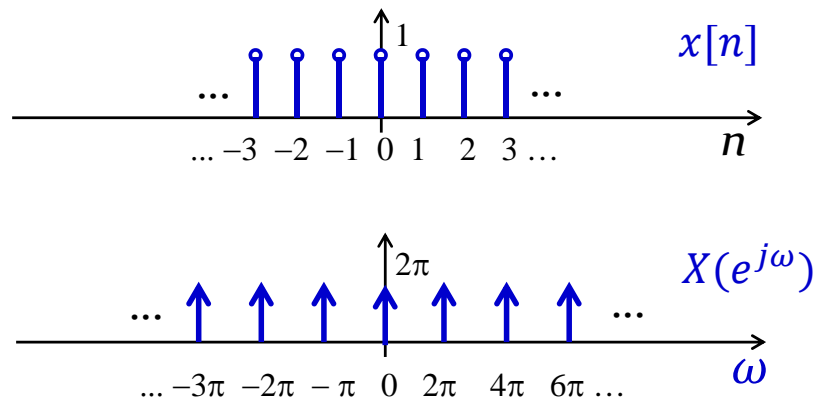


- **Gibbs phenomenon:** converges at each ω as $K \rightarrow \infty$, but peak error does not get smaller.

- The DTFT can also be defined for a certain class of sequences which are neither absolutely summable nor square summable
- Examples of such sequences are the unit step sequence $\mu[n]$ and the sinusoidal sequence $\cos(\omega_0 n + \varphi)$.
- For this type of sequences, a DTFT representation is possible using the Dirac delta function $\delta[\omega]$.

Example 5: DTFT of $x[n]=1$ for all n

- $x[n] = 1 = \sum_{k=-\infty}^{\infty} \delta[n - k]$
- It is more convenient to **prove that the inverse DTFT of $\sum_{k=-\infty}^{\infty} 2\pi\delta(\omega - 2\pi k)$ is 1**



- *Proof:* The inverse DTFT of $\sum_{k=-\infty}^{\infty} 2\pi\delta(\omega - 2\pi k)$ is evaluated as

$$\begin{aligned} & \frac{1}{2\pi} \int_{-\pi}^{\pi} \left(\sum_{k=-\infty}^{\infty} 2\pi\delta(\omega - 2\pi k) \right) e^{j\omega n} d\omega \\ &= \int_{-\pi}^{\pi} \left(\sum_{k=-\infty}^{\infty} \delta(\omega - 2\pi k) \right) e^{j\omega n} d\omega \end{aligned}$$

- From the sifting property, we have

$$\begin{aligned} \left(\sum_{k=-\infty}^{\infty} \delta(\omega - 2\pi k) \right) e^{j\omega n} &= \left(\sum_{k=-\infty}^{\infty} \delta(\omega - 2\pi k) \right) e^{j2\pi kn} \\ &= \sum_{k=-\infty}^{\infty} \delta(\omega - 2\pi k) \end{aligned}$$

We have used $e^{j2\pi kn} = 1$ for all n here

- When we integrate the sequence of impulse from $-\pi$ to π , we have only the impulse at $\omega = 0$.
- Hence

$$\begin{aligned}
 & \int_{-\pi}^{\pi} \left(\sum_{k=-\infty}^{\infty} \delta(\omega - 2\pi k) \right) e^{j\omega n} d\omega \\
 &= \int_{-\pi}^{\pi} \left(\sum_{k=-\infty}^{\infty} \delta(\omega - 2\pi k) \right) d\omega \\
 &= \int_{-\pi}^{\pi} \delta(\omega) d\omega = \int_{-\infty}^{-\pi} \delta(\omega) d\omega + \int_{-\pi}^{\pi} \delta(\omega) d\omega + \int_{\pi}^{\infty} \delta(\omega) d\omega \\
 &= \int_{-\infty}^{\infty} \delta(\omega) d\omega = 1 \quad \text{for all } n
 \end{aligned}$$

Example 6: DTFT of $\mu[n]$

- Let $\mu[n] = u_1[n] + u_2[n]$, where

$$u_1[n] = \frac{1}{2}, \quad \text{for } -\infty < n < \infty$$

and

$$u_2[n] = \begin{cases} \frac{1}{2} & \text{for } n \geq 0 \\ -\frac{1}{2} & \text{for } n < 0 \end{cases}$$

Therefore, we have

$$\delta[n] = u_2[n] - u_2[n-1]$$

- Using $\delta[n] \leftrightarrow 1$ and $u_2[n] - u_2[n-1] \leftrightarrow U_2(e^{j\omega}) - e^{-j\omega}U_2(e^{j\omega}) = U_2(e^{j\omega})(1 - e^{-j\omega})$, we have

$$1 = U_2(e^{j\omega})(1 - e^{-j\omega})$$

i.e.

$$U_2(e^{j\omega}) = \frac{1}{1 - e^{-j\omega}}$$

Since

$$u_1[n] \leftrightarrow \sum_{k=-\infty}^{\infty} \pi\delta(\omega - 2\pi k) = U_1(e^{j\omega})$$

we have

$$\mu[n] \leftrightarrow \frac{1}{1 - e^{-j\omega}} + \sum_{k=-\infty}^{\infty} \pi\delta(\omega + 2\pi k)$$

DTFT Convergence

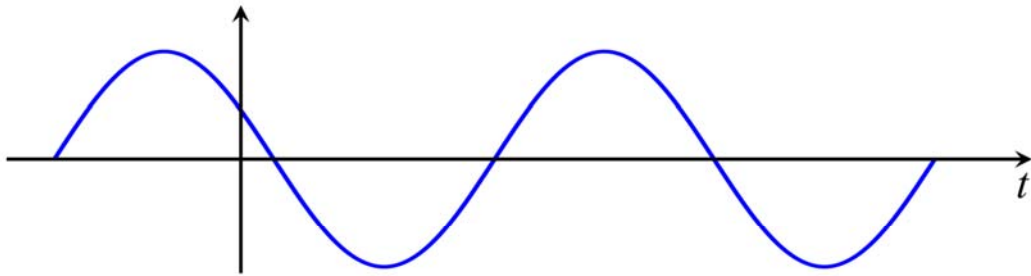
Sequence		DTFT
$\alpha^n \mu[n], (\alpha < 1)$ Absolutely Summable	Sufficient \rightarrow	$\frac{1}{1 - \alpha e^{-j\omega}}$ Exist for all ω
$\mu[n]$ Neither absolutely summable, nor finite energy	Not necessary \rightarrow	$\frac{1}{1 - e^{-j\omega}} + \sum_{k=-\infty}^{\infty} \pi\delta(\omega + 2\pi k)$ Not exist for $\omega = 0$
1 (for all n) Neither absolutely summable, nor finite energy	Not necessary \rightarrow	$\sum_{k=-\infty}^{\infty} 2\pi\delta(\omega - 2\pi k)$ Exist for all ω
$h_{LP}[n] = \frac{\sin\omega_c n}{\pi n}$ Finite energy	Sufficient \rightarrow	$H_{LP}(e^{j\omega}) = \begin{cases} 1, & 0 \leq \omega \leq \omega_c\pi \\ 0, & \omega_c\pi \leq \omega \leq \pi \end{cases}$ Exist for all ω

Commonly used DTFT pairs

Sequence	DTFT
$\delta[n]$	1
$\mu[n]$	$\frac{1}{1 - e^{-j\omega}} + \sum_{k=-\infty}^{\infty} \pi \delta(\omega + 2\pi k)$
1 (for all n)	$\sum_{k=-\infty}^{\infty} 2\pi \delta(\omega - 2\pi k)$
$\alpha^n \mu[n], (\alpha < 1)$	$\frac{1}{1 - \alpha e^{-j\omega}}$

Effect of Time-Domain Sampling in Frequency Domain

- Questions to be answered
 - When discrete signal is obtained by sampling, can we recover the original continuous signal from the discrete signal?
 - What is the condition that we can recover the continuous signal?
 - Relation Between Continuous and Discrete Signal Spectrum



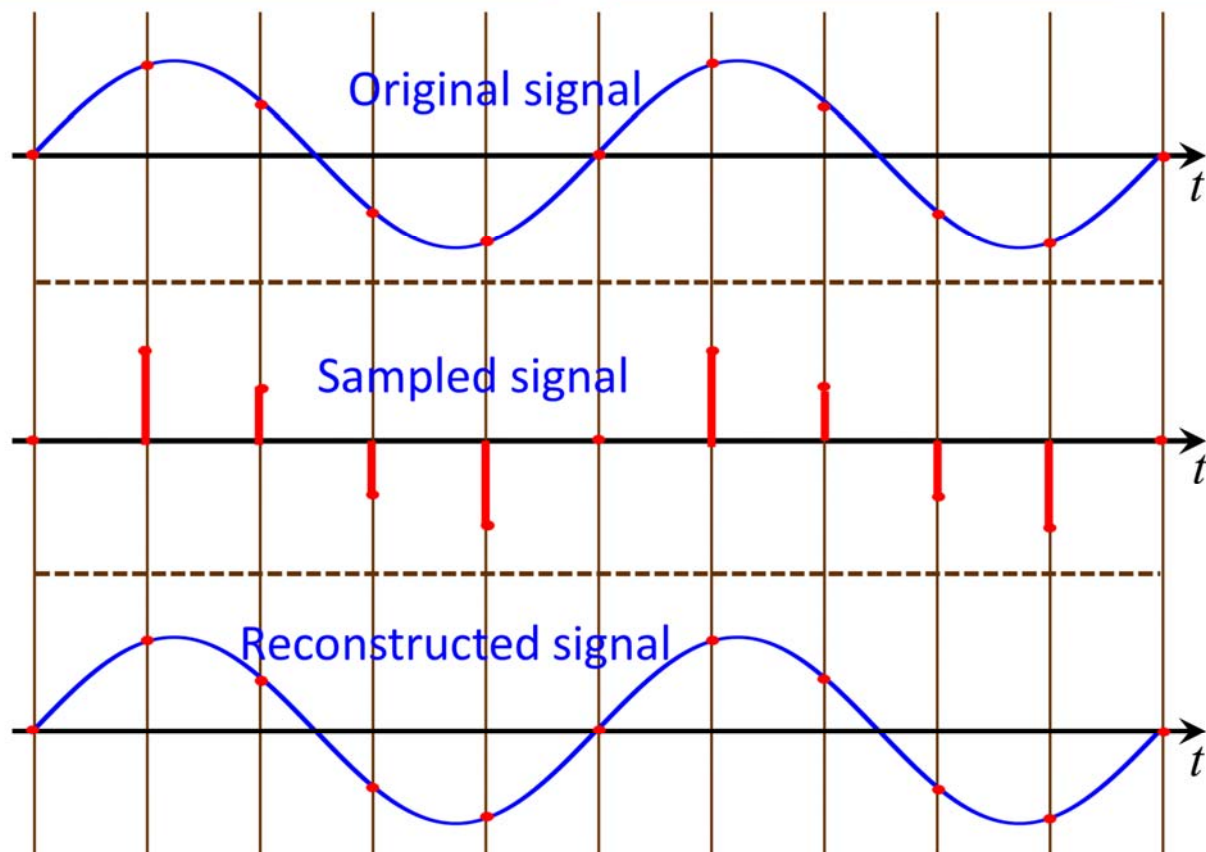
A continuous time signal may be sampled to produce a discrete time signal.



If the rate of sampling is very high, it is obvious by inspection that the sampled result will resemble that of the original continuous time signal.

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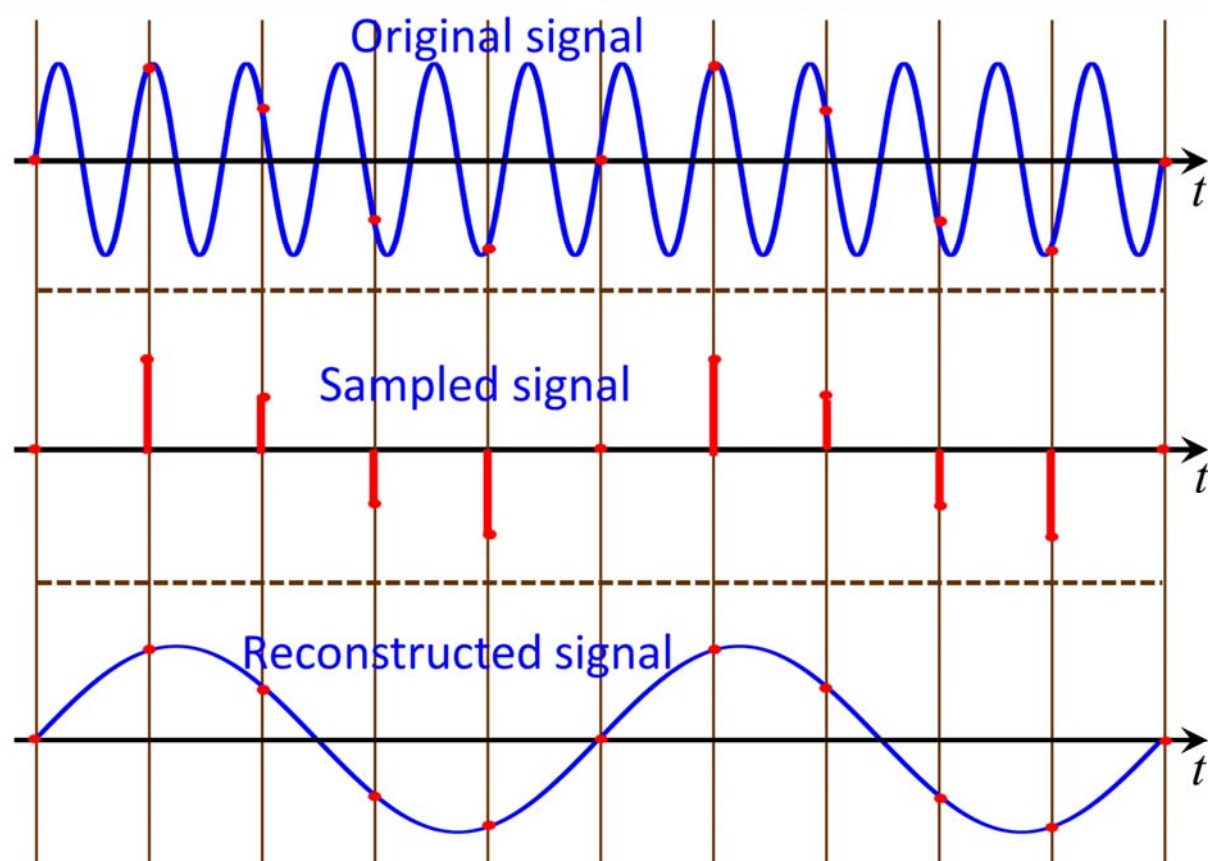
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If the sampling rate is sufficiently high, it is possible to reconstruct the original signal from the sampled signal.

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If the sampling rate is not sufficiently high, the reconstructed signal is different from the original signal.

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Fourier transform of a discrete time signal.

Let $x_c(t)$ be a continuous time signal.

Let the sampled version of $x_c(t)$ be denoted by $x_d(t)$.

Let the sampling interval be T .

Let $\omega_s = 2\pi/T$.

Let $X_c(j\omega)$ be the Fourier transform of $x_c(t)$.

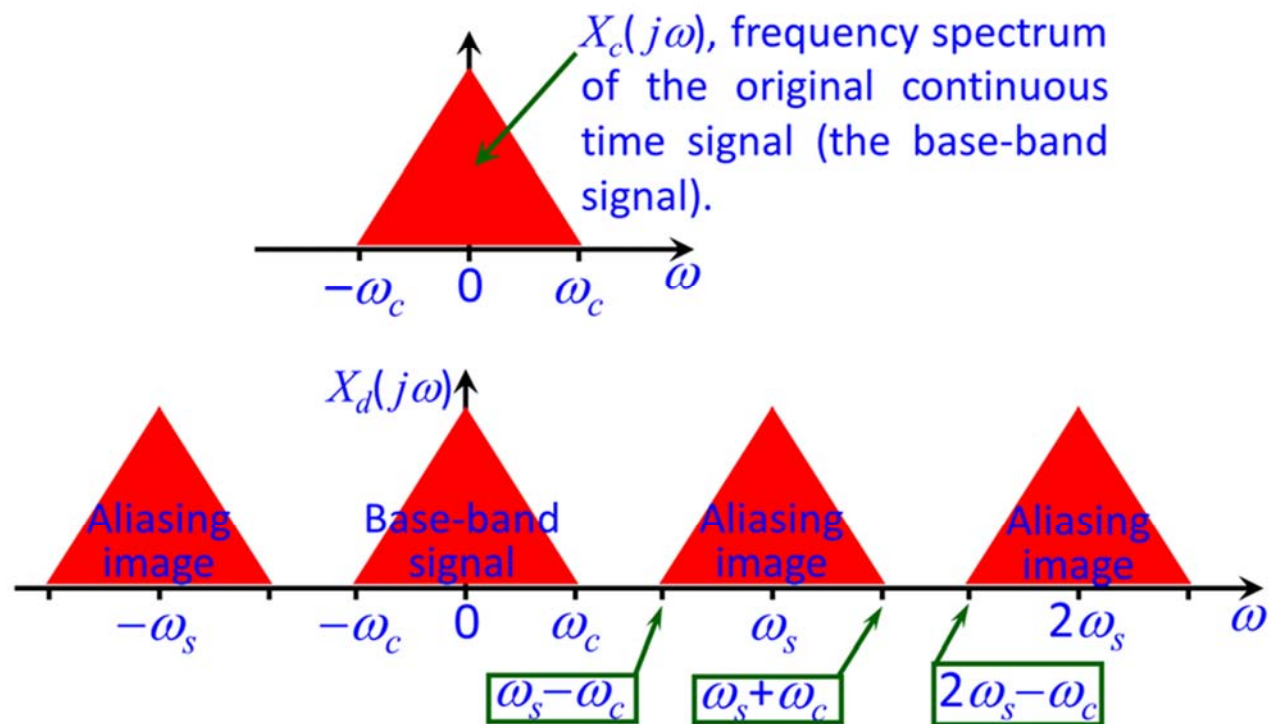
Let $X_d(j\omega)$ be the Fourier transform of $x_d(t)$.

It can be shown that
$$X_d(j\omega) = \frac{1}{T} \sum_{n=-\infty}^{\infty} X_c(j(\omega - n\omega_s))$$

$1/T$ indicates that the magnitude of $X_d(j\omega)$ increases with sampling density $1/T$.

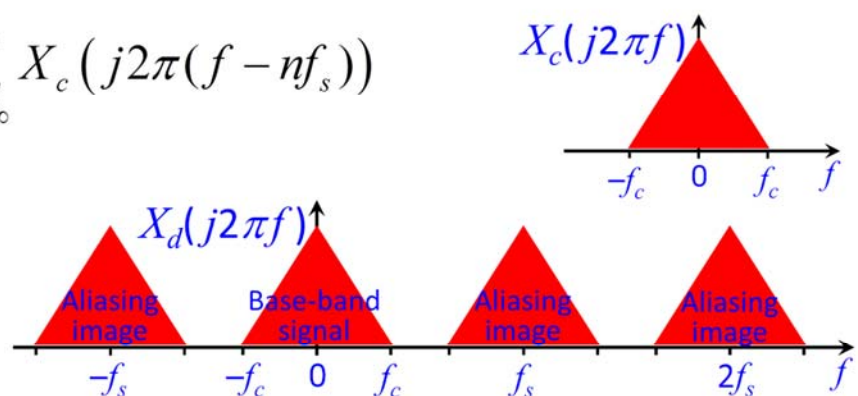
$X_c(j(\omega - n\omega_s))$ is $X_c(j\omega)$ shifted along the ω -axis by $n\omega_s$.

$$X_d(j\omega) = \frac{1}{T} \sum_{n=-\infty}^{\infty} X_c(j(\omega - n\omega_s))$$



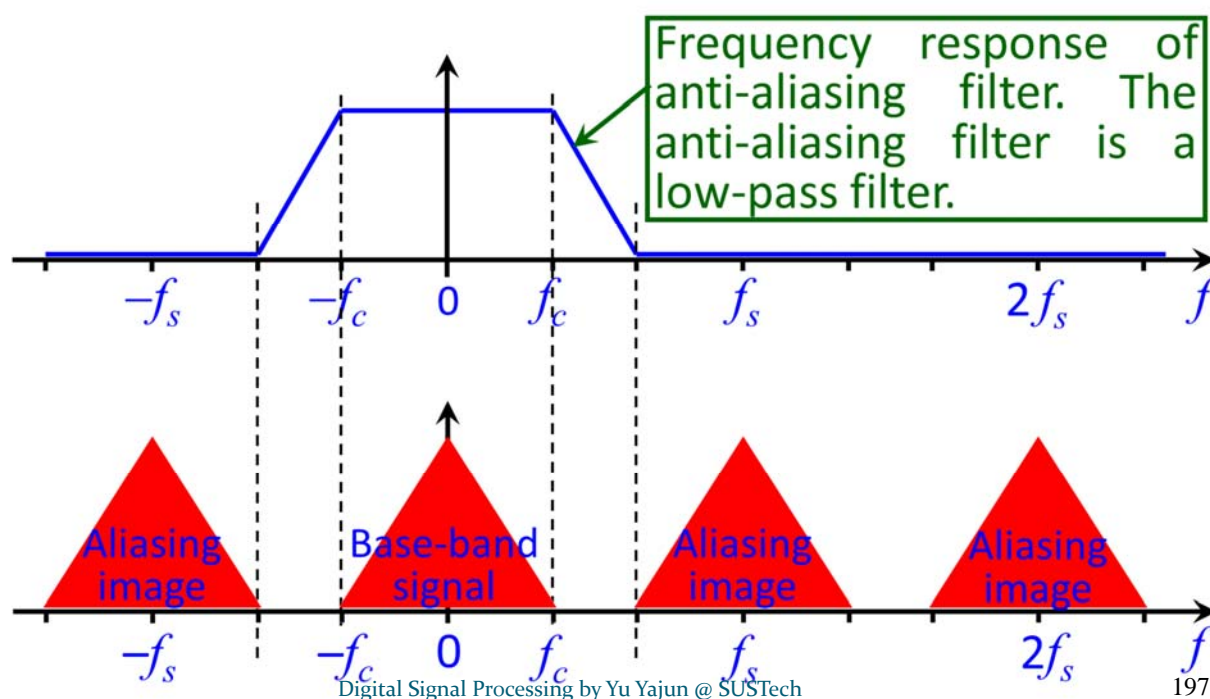
“Hz” is often used as frequency unit in communication systems. Hence, replacing ω by $2\pi f$ we have

$$X_d(j2\pi f) = \frac{1}{T} \sum_{n=-\infty}^{\infty} X_c(j2\pi(f - nf_s))$$

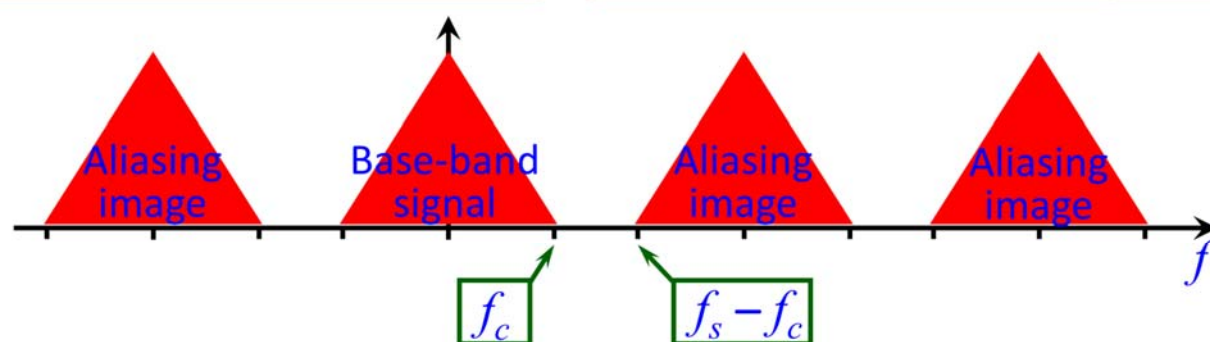


When a continuous time signal (the base-band signal) is sampled at a rate of f_s samples per second, the frequency spectrum of the sampled signal is that of the base-band signal plus duplicates (aliasing) of that of the base-band signal centred at kf_s where $k = \dots, -1, 0, 1, 2, 3, \dots$

The original continuous time signal can be recovered by removing the aliasing images by low-pass filtering. The low-pass filter is called an anti-aliasing filter.



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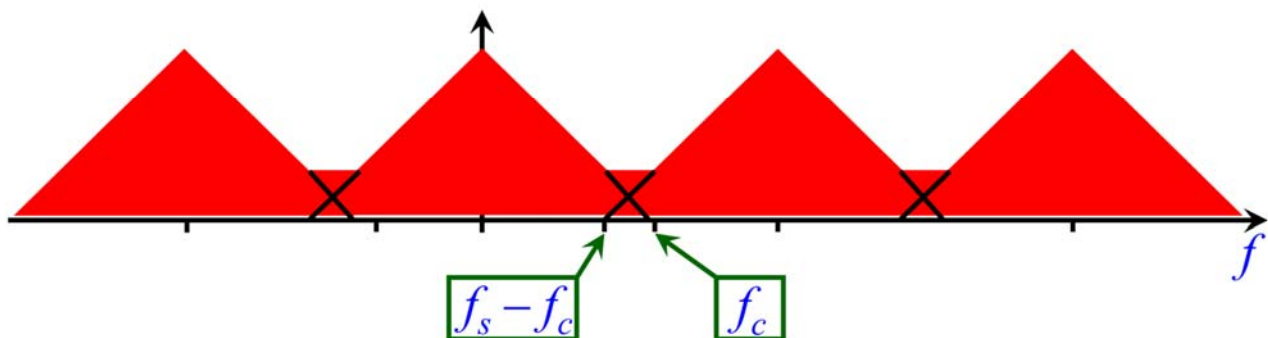
If the original continuous time signal can be completely recovered, the aliasing is not a serious problem. Recovering of the original continuous time signal is possible if and only if there is no overlap between the base-band signal and the aliasing images.

$$f_s - f_c > f_c, \text{ i.e. } f_s > 2f_c.$$

Nyquist Frequency & Nyquist Rate

- The highest frequency (f_c) contained in a continuous signal $x_c(t)$ is usually called **Nyquist frequency**, which determines the minimum sampling frequency ($f_s = 2f_c$) that must be used to fully recover $x_c(t)$ from its sampled version.
- The minimum sampling frequency (or sampling rate), $f_s = 2f_c$, required to avoid aliasing from irrecoverable problem is called **Nyquist rate**.

If $f_s - f_c < f_c$, i.e. $f_s < 2f_c$,

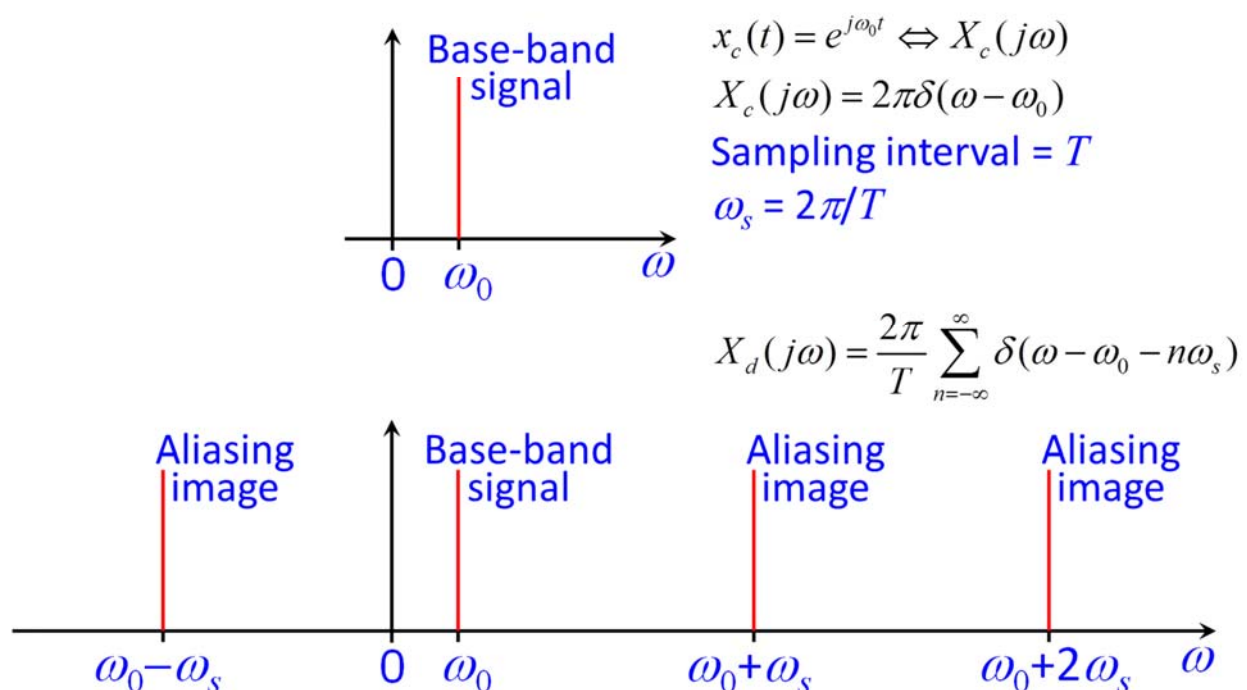


Example:

The output of an ideal low-pass filter with band-edges at $\pm\pi$ radians per second is sampled at a rate of f_s samples per second. What is the minimum f_s in order to avoid aliasing from causing irrecoverable problem?

Since the band-edges of the ideal low-pass filter is $\pm\pi$ radians per second, the highest frequency component of the low-pass filter output is π radians per second. We have $\omega = 2\pi f$. Thus, the highest frequency component of the low-pass filter output is $\pi/(2\pi) = 1/2$ Hz. The sampling rate must be greater than twice the maximum frequency. Hence, minimum sampling rate is 1 sample per second.

Spectral lines of a sampled complex sinusoid.



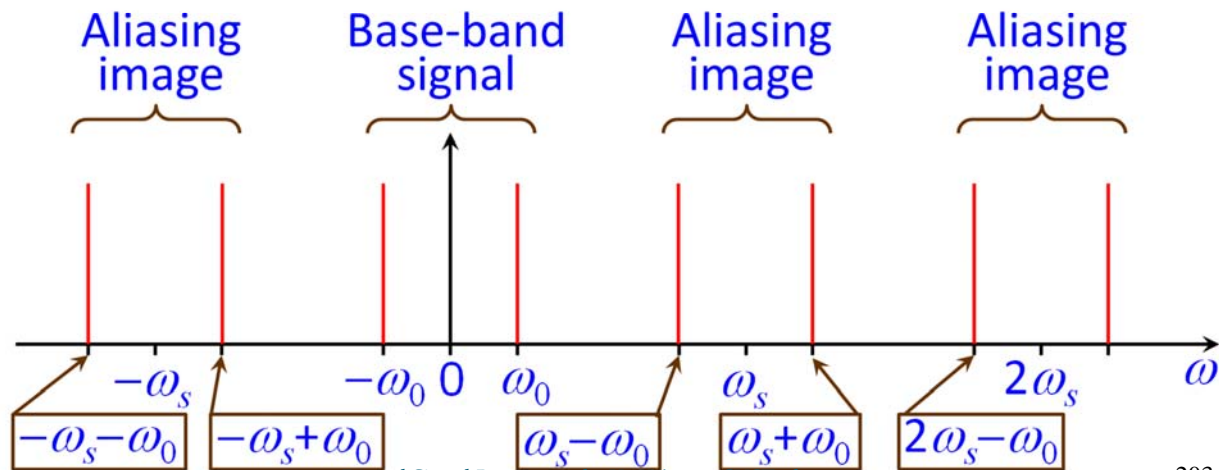
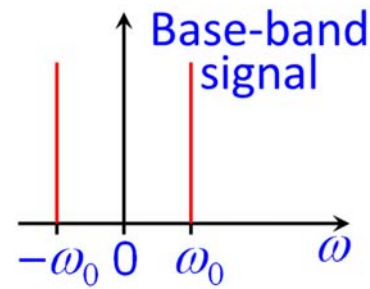
Spectral lines of a sampled sinusoid.

$$\cos(\omega_0 t) = \frac{e^{j\omega_0 t} + e^{-j\omega_0 t}}{2}$$

Sampling interval = T

$$\omega_s = 2\pi/T$$

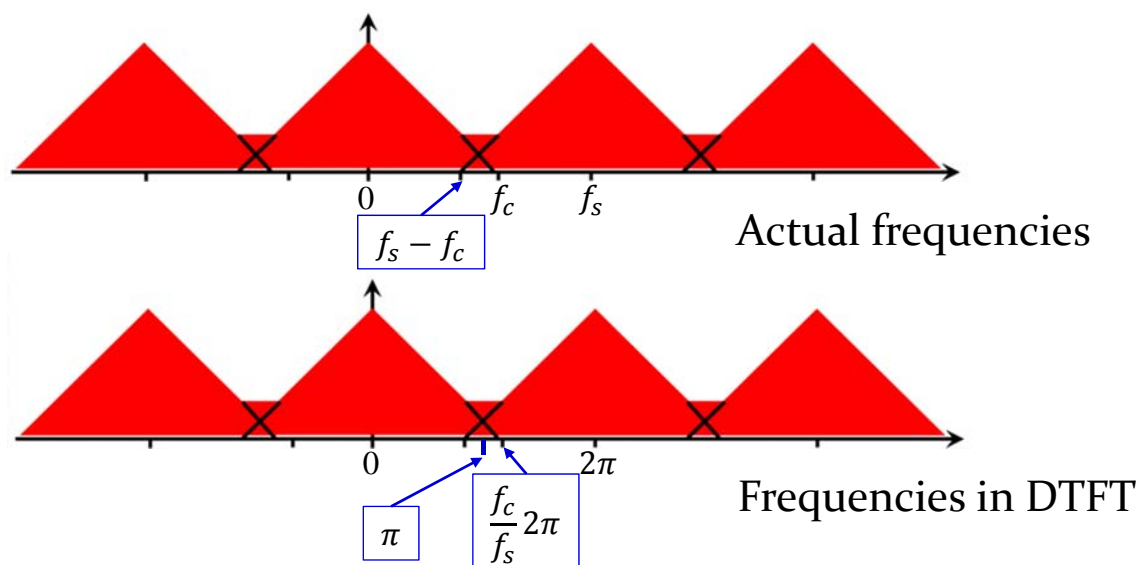
$$X_d(j\omega) = \frac{\pi}{T} \sum_{n=-\infty}^{\infty} \delta(\omega \pm \omega_0 - n\omega_s)$$



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Normalization of frequency in DTFT

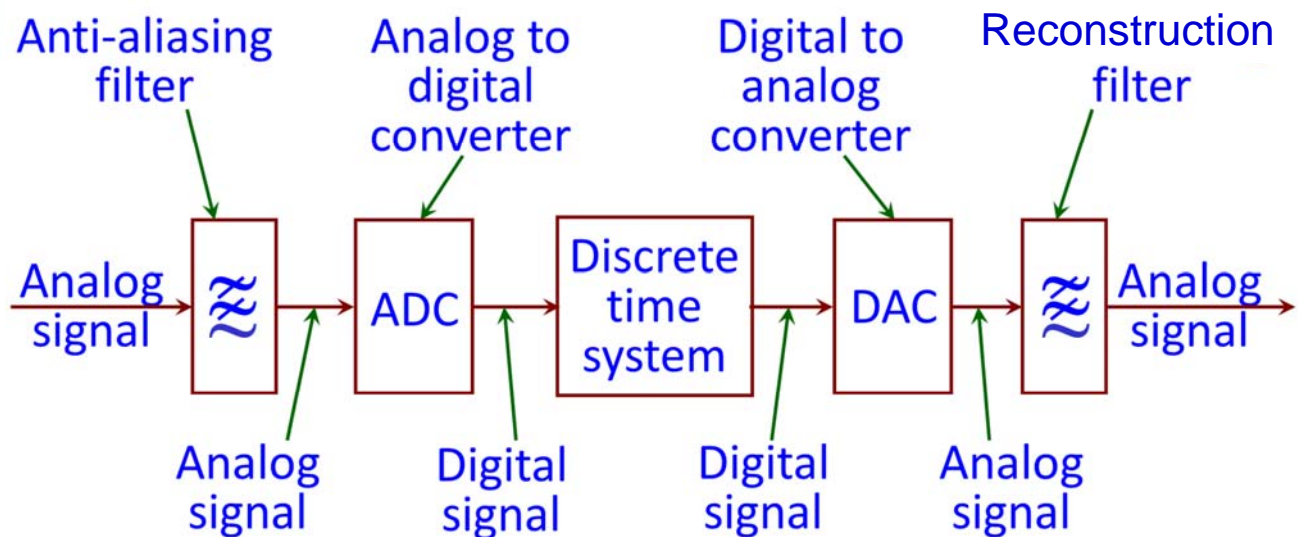


- Actual sampling frequency $f_s \leftrightarrow 2\pi$ in DTFT
- Normalization of frequency: $f \leftrightarrow \frac{f}{f_s} 2\pi$

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A typical discrete time system configuration



Examples of sampling rate:

(a) CD : 44.1 kHz.

(b) Digital audio tape : 48 kHz.

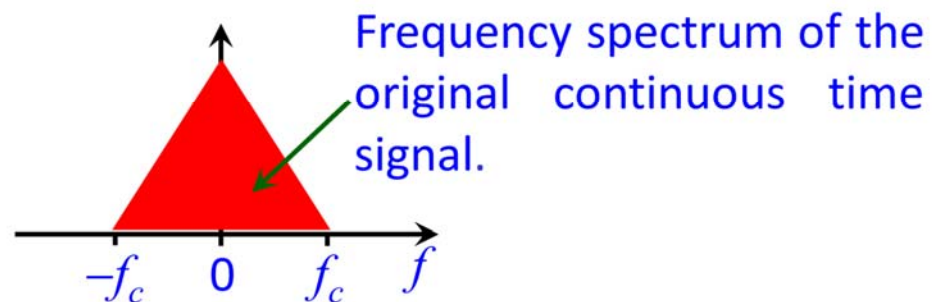
(c) Telephone system : 8 KHz.

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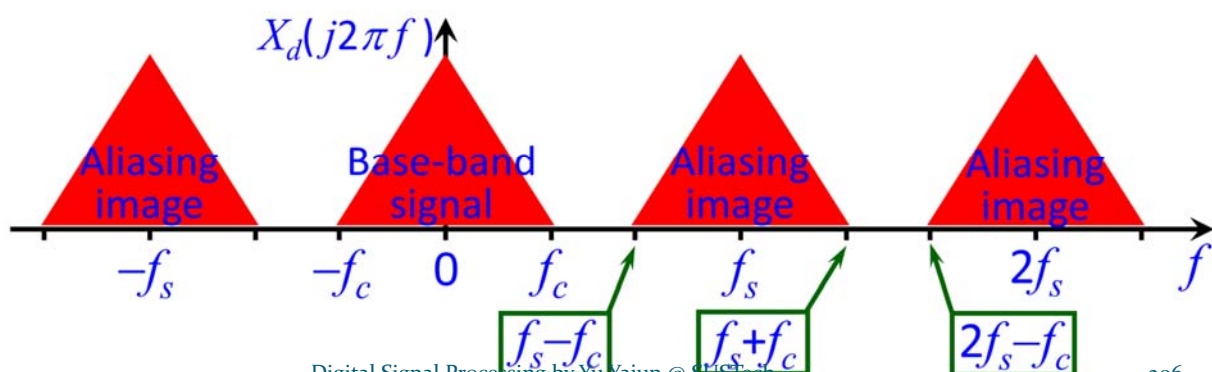
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Summary:

1.



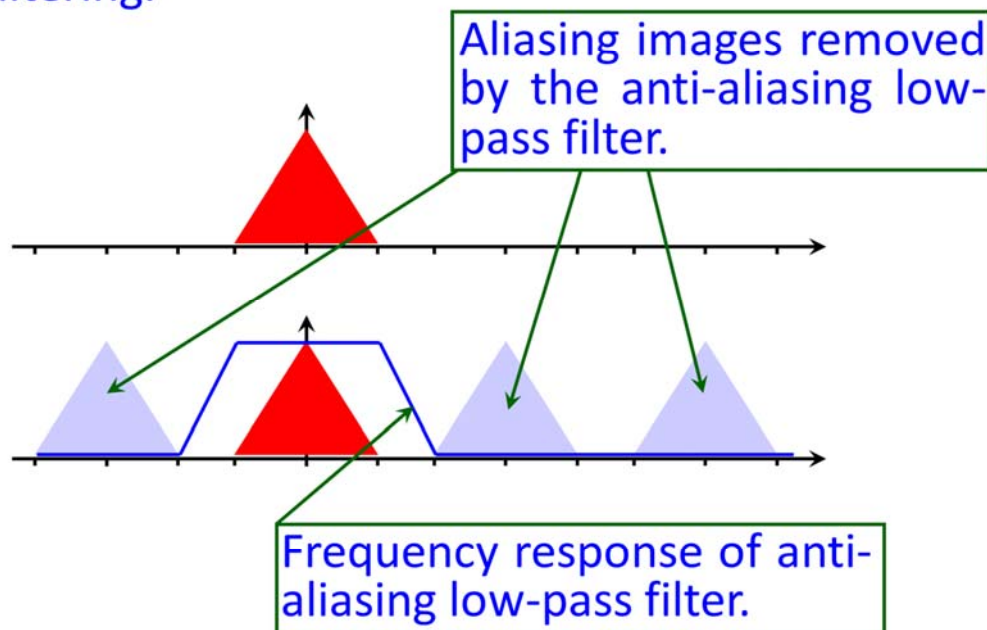
Frequency spectrum of the sampled discrete time signal.



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- Information in the original continuous time signal can be recovered from the sampled signal by low-pass filtering.



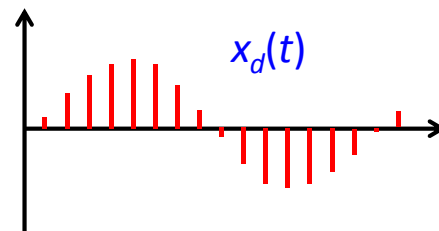
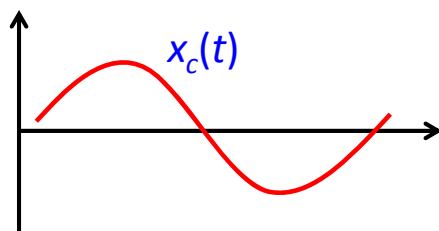
- Nyquist rate = $2 \times$ maximum frequency.

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Mathematical Derivation

To prove
$$X_d(j\omega) = \frac{1}{T} \sum_{n=-\infty}^{\infty} X_c(j(\omega - n\omega_s))$$



we model

$$x_d(t) = x_c(t) \sum_{n=-\infty}^{\infty} \delta(t - nT), \text{ where } \delta(t) = \begin{cases} \infty & \text{for } t = 0 \\ 0 & \text{for } t \neq 0 \end{cases}$$

$$\int_{-\infty}^{\infty} \delta(t) dt = 1$$

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Recall, the Fourier Series of

$$\sum_{n=-\infty}^{\infty} \delta(t - nT) = \frac{1}{T} \sum_{n=-\infty}^{\infty} e^{jn\omega_s}, \quad \text{where } \omega_s = \frac{2\pi}{T}$$

Thus,

$$x_d(t) = x_c(t) \frac{1}{T} \sum_{n=-\infty}^{\infty} e^{jn\omega_s t},$$



Now, look at the spectrum of the transformed signal.
Using the convolution property, we have

$$\begin{aligned} X_d(j\omega) &= \frac{1}{2\pi} \frac{1}{T} X_c(j\omega) \otimes \sum_{n=-\infty}^{\infty} 2\pi \delta(\omega - n\omega_s) \\ &= \frac{1}{T} \int_{-\infty}^{\infty} X_c(j\varphi) \sum_{n=-\infty}^{\infty} 2\pi \delta(\omega - n\omega_s - \varphi) d\varphi \\ &= \frac{1}{T} \sum_{n=-\infty}^{\infty} X_c(j(\omega - n\omega_s)) \end{aligned}$$