# Lecture 12: Sampling, Aliasing, and the Discrete Fourier Transform

Foundations of Digital Signal Processing

#### **Outline**

- Review of Sampling
- The Nyquist-Shannon Sampling Theorem
- Continuous-time Reconstruction / Interpolation
- Aliasing and anti-Aliasing
- Deriving Transforms from the Fourier Transform
  - Discrete-time Fourier Transform, Fourier Series, Discrete-time Fourier Series
- The Discrete Fourier Transform

#### News

- Homework #5
  - Due <u>this week</u>
  - Submit via canvas
- Coding Problem #4
  - Due <u>this week</u>
  - Submit via canvas

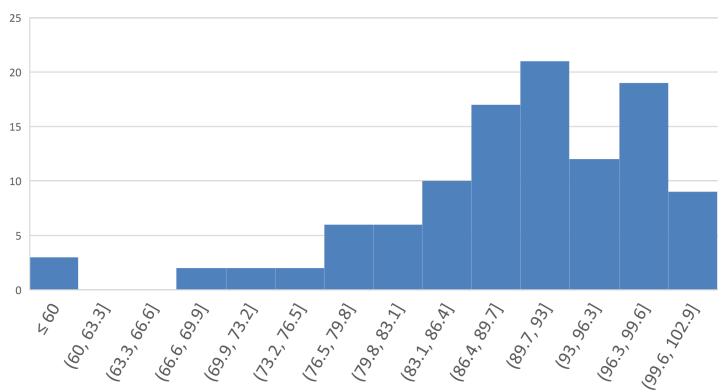
#### Exam 1 Grades

#### The class did exceedingly well

Mean: 89.3

Median: 91.5





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Discrete-Time Fourier Transform

#### Discrete-Time Fourier Transform

Example: Consider the DTFT signal

$$X(\omega) = e^{-\frac{j\omega\pi}{8}} \sum_{k=-\infty}^{\infty} u\left(\omega + \frac{\pi}{2} - 2\pi k\right) - u\left(\omega + \frac{\pi}{4} - 2\pi k\right) + u\left(\omega - \frac{\pi}{2} - 2\pi k\right) - u\left(\omega - \frac{\pi}{4} - 2\pi k\right)$$

• Sketch the magnitude of  $X(\omega)$ .

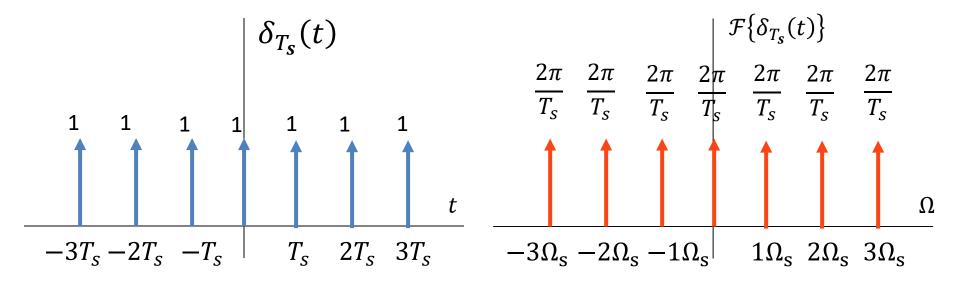
#### Discrete-Time Fourier Transform

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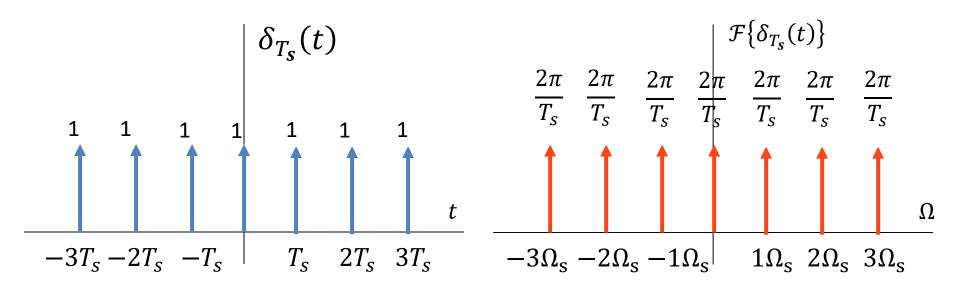
• Sketch the phase of  $X(\omega)$ .

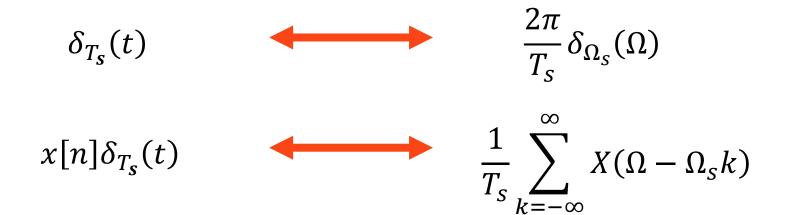
$$\delta_{T_s}(t)$$
 
$$\frac{2\pi}{T_s}\delta_{\Omega_s}(\Omega)$$

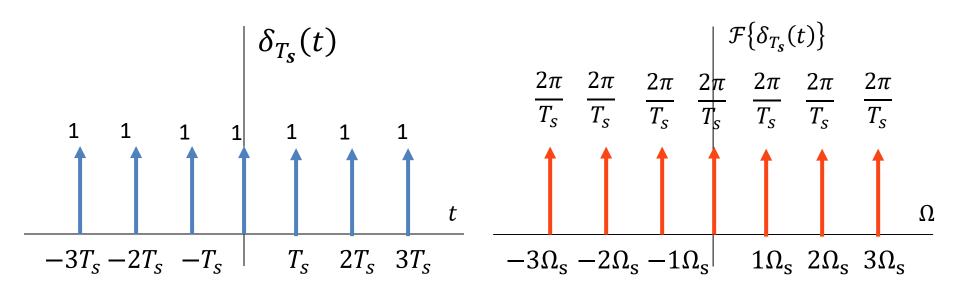


$$\delta_{T_s}(t) \qquad \frac{2\pi}{T_s} \delta_{\Omega_s}(\Omega)$$

$$x[n]\delta_{T_s}(t) \qquad \frac{1}{2\pi} X(\Omega) * \left[\frac{2\pi}{T_s} \delta_{\Omega_s}(\Omega)\right]$$







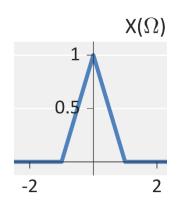
# Lecture 12: Sampling, Aliasing, and the Discrete Fourier Transform

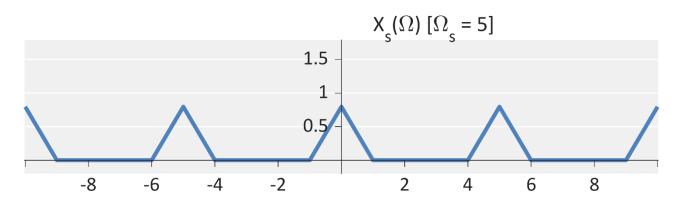
Foundations of Digital Signal Processing

#### **Outline**

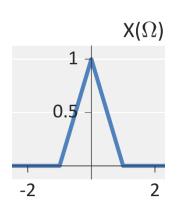
- Review of Sampling
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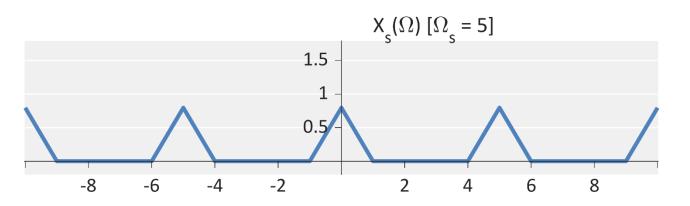
- Question: Can I preserve all information when I sample?
  - Yes!

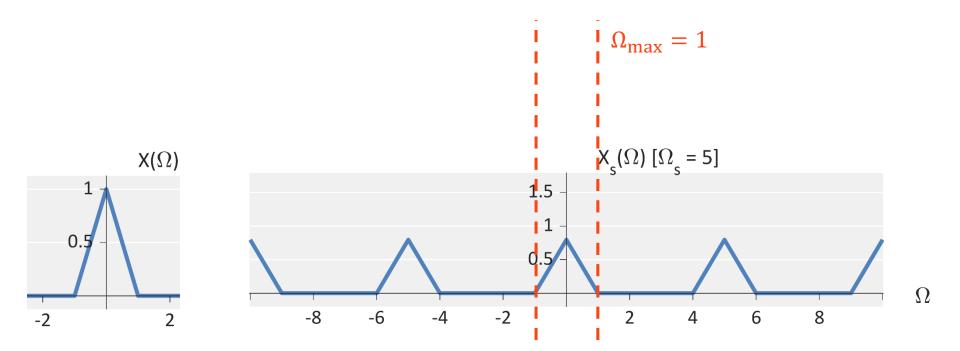


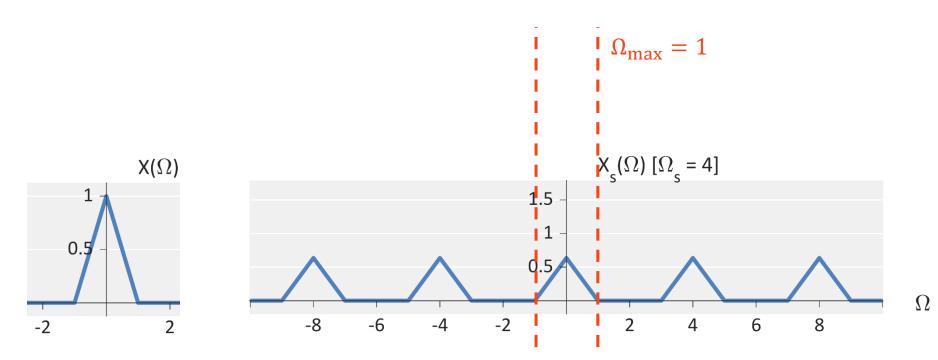


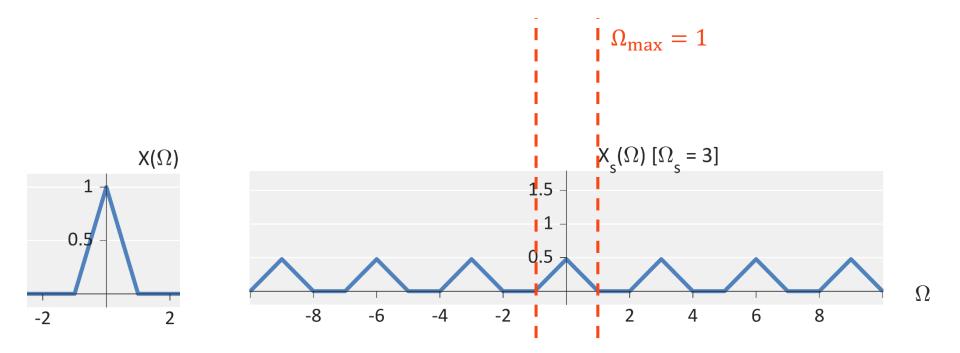
Question: How fast do I sample to preserve information?

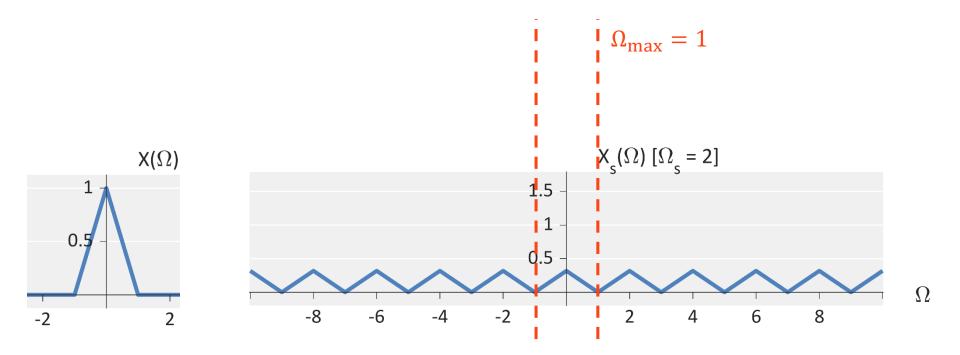






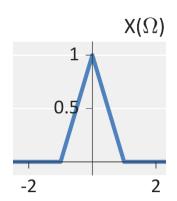


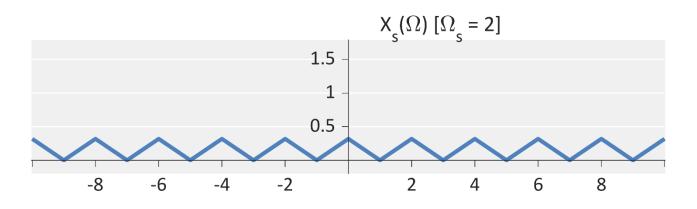




- Question: How fast do I sample to preserve information?
  - We need to sample twice as fast as the maximum frequency

$$\Omega_{\rm s} > 2\Omega_{\rm max}$$

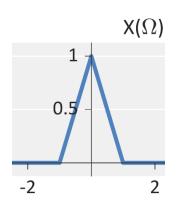


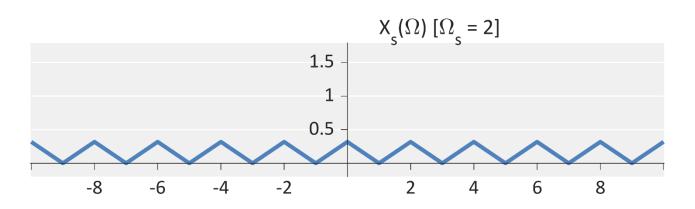


#### Question: How fast do I sample to preserve information?

We need to sample twice as fast as the maximum frequency

$$\begin{split} &\Omega_{\text{S}} > 2\Omega_{\text{max}} & \text{Nyquist-Shannon} \\ &f_{\text{S}} > 2f_{\text{max}} & \text{Sampling Theorem} \end{split}$$

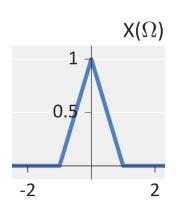


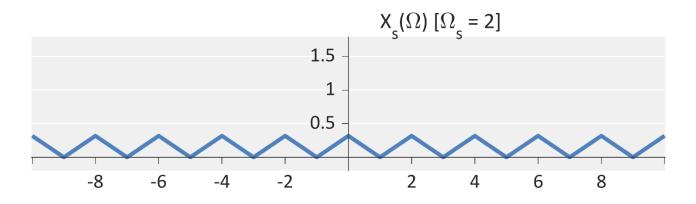


#### Question: How fast do I sample to preserve information?

We need to sample twice as fast as the maximum frequency

$$\Omega_{s}>2\Omega_{max}$$
 Nyquist-Shannon 
$$f_{s}>2f_{max}$$
 Sampling Theorem 
$$\Omega_{N}=2\Omega_{max}<\text{--Nyquist Rate}$$





#### Example

- **Example: Consider the signal**  $x(t) = \cos(5\pi t)$ .
  - What is the Nyquist rate?
  - Sketch the Fourier transform  $X_S(\Omega)$  of the sampled signal when  $\Omega_S=12\pi$

#### Example

**Example:** Consider the frequency signal

$$X(\Omega) = u(\omega + 5\pi) - u(\omega - 5\pi)$$

- What is the Nyquist rate?
- Sketch the Fourier transform  $X_{S}(\Omega)$  of the sampled signal when  $\Omega_{\rm S}=20\pi$

# Lecture 12: Sampling, Aliasing, and the Discrete Fourier Transform

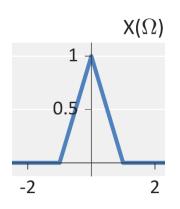
Foundations of Digital Signal Processing

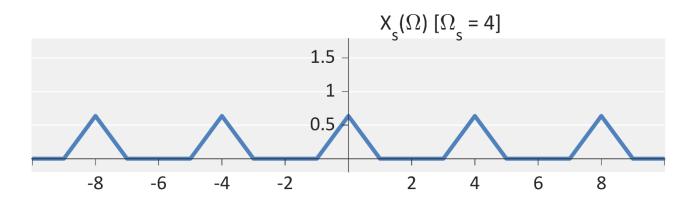
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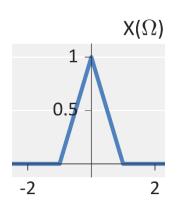
#### Question: How do I return to continuous—time?

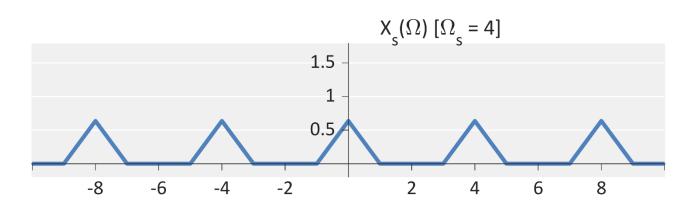






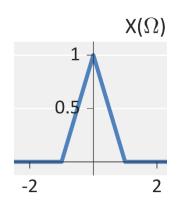
- Question: How do I return to continuous—time?
  - Filter: Low pass filter to keep  $\Omega_s/2 \leq \Omega \leq \Omega_s/2$
  - Amplify: Amplify signal by  $T_S$

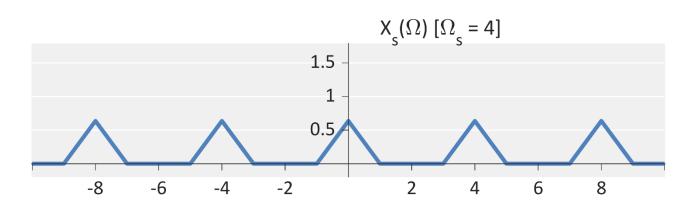




#### Question: How do I return to continuous—time?

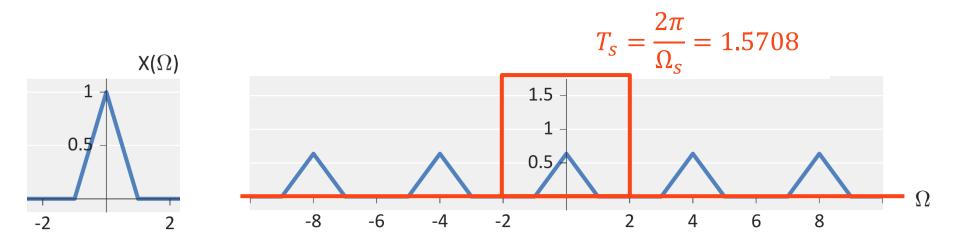
- Apply a low-pass reconstruction filter
  - Cut-off frequency:  $\Omega_s/2$
  - Gain:  $T_S$





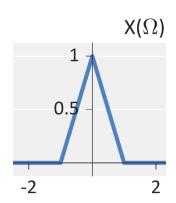
#### Question: How do I return to continuous—time?

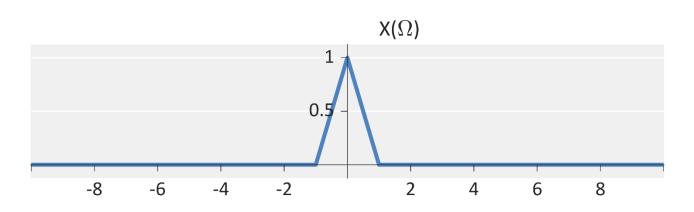
- Apply a low-pass reconstruction filter
  - Cut-off frequency:  $\Omega_s/2$
  - $\diamond$  Gain:  $T_S$



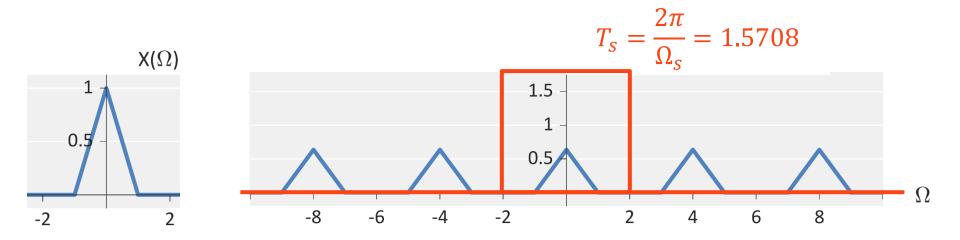
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- Apply a low-pass reconstruction filter
  - Cut-off frequency:  $\Omega_s/2$
  - $\diamond$  Gain:  $T_S$

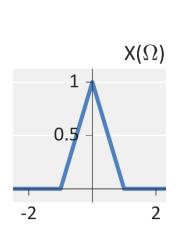


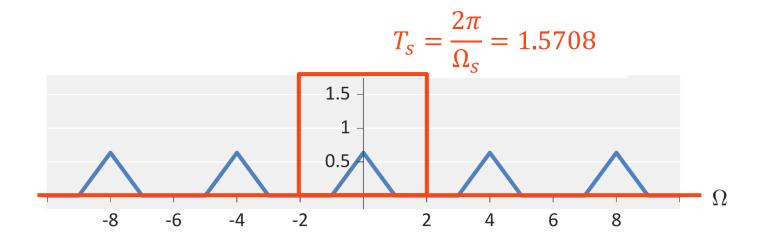


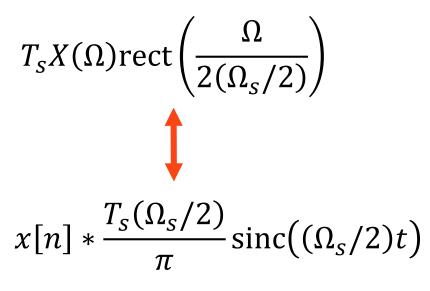
$$X(\Omega)[T_s[u(\Omega + \Omega_s/2) - u(\Omega - \Omega_s/2)]]$$

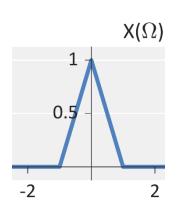


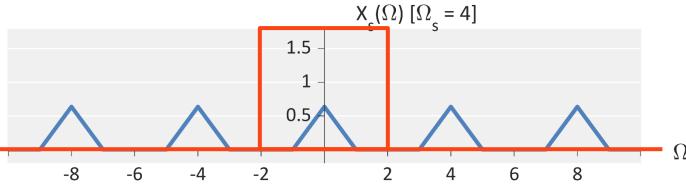
$$X(\Omega)\left[T_SX(\Omega)\operatorname{rect}\left(\frac{\Omega}{\Omega_S}\right)\right]$$





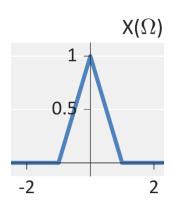


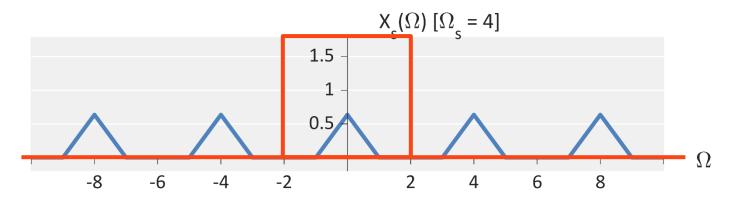




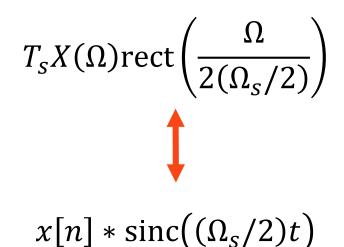
$$T_s X(\Omega) \operatorname{rect}\left(\frac{\Omega}{2(\Omega_s/2)}\right)$$

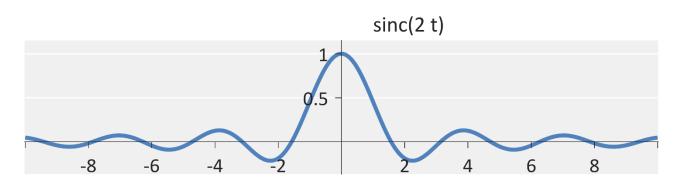
$$x[n] * \frac{T_S(\Omega_S/2)}{\pi} \operatorname{sinc}((\Omega_S/2)t) = x[n] * \operatorname{sinc}((\Omega_S/2)t)$$





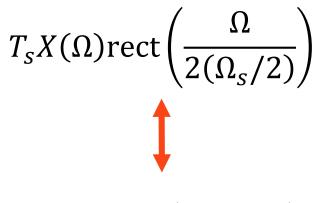
Question: What is happening when I multiply in frequency?



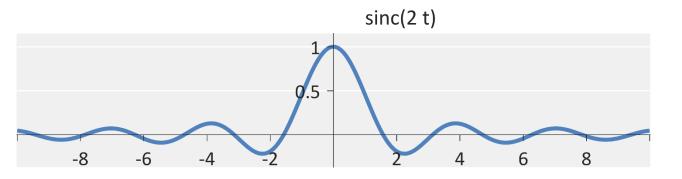


t

Question: What is happening when I multiply in frequency?



$$x[n] * \operatorname{sinc}((\Omega_s/2)t)$$

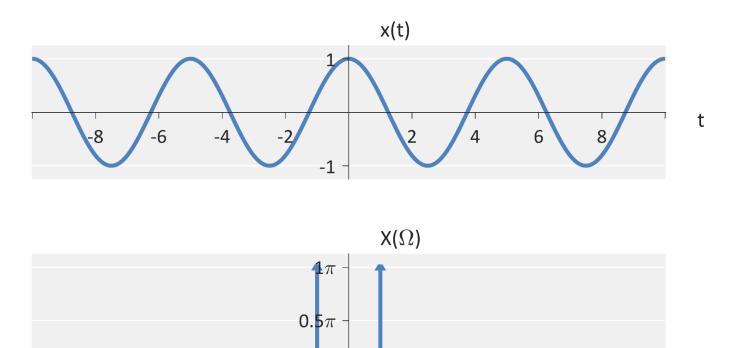


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#### Reconstruction

#### Consider a cosine

-10

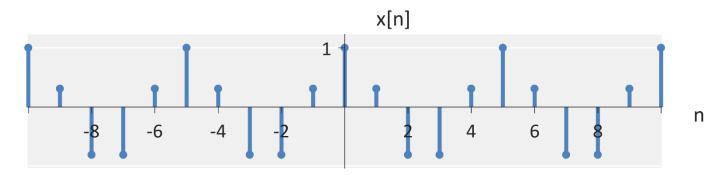


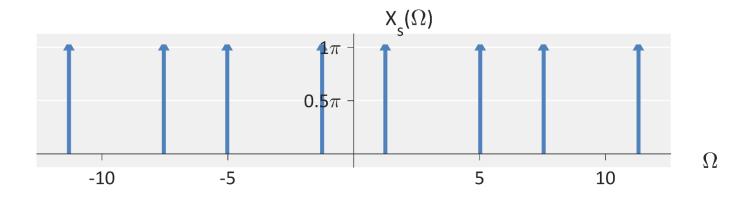
5

-5

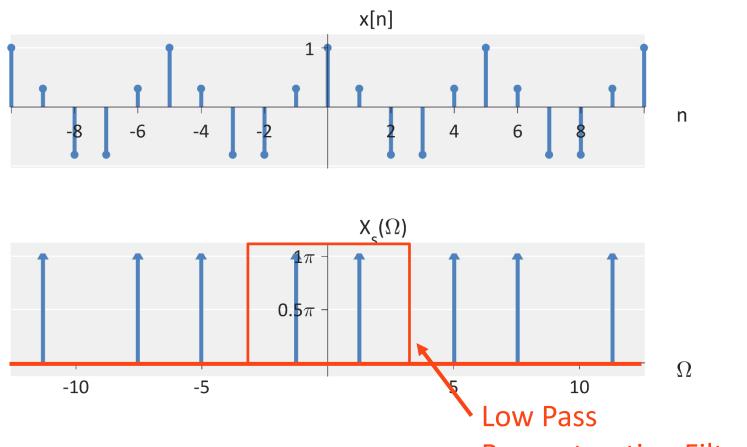
10

#### ■ Consider a cosine – Sampled at $T_s = 1$

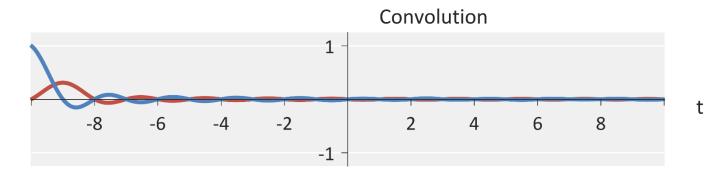


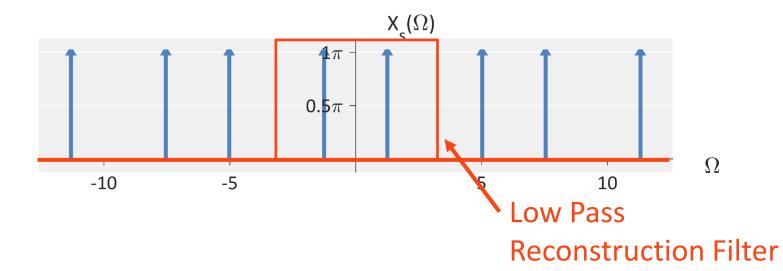


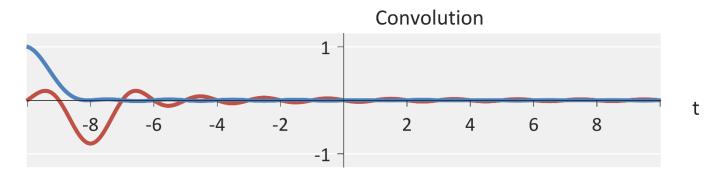
■ Consider a cosine – Low pass filter to keep  $-\Omega_s/2$  to  $\Omega_s/2$ 

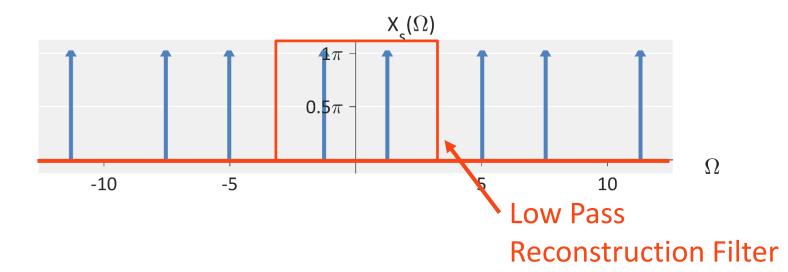


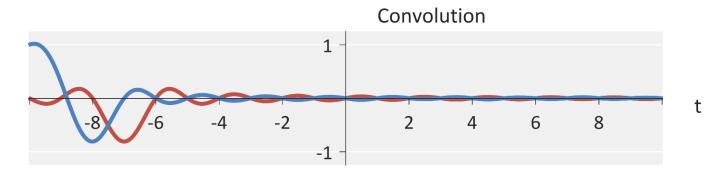
**Reconstruction Filter** 

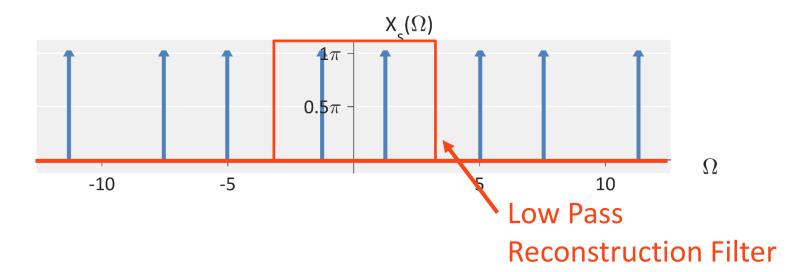


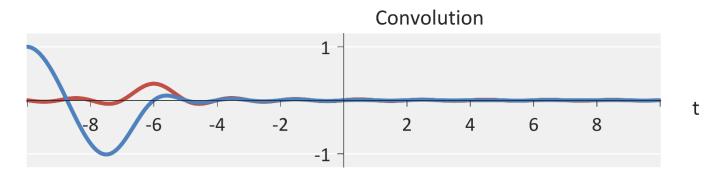


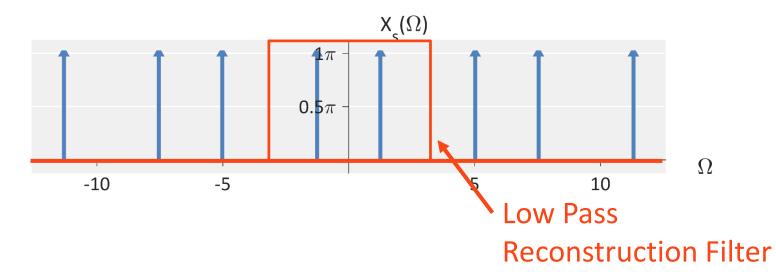


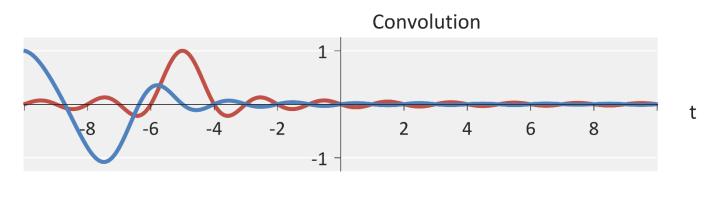


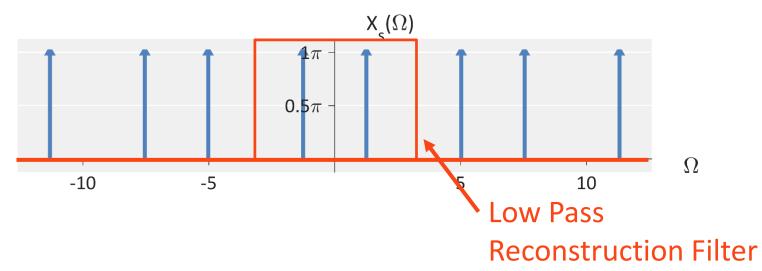


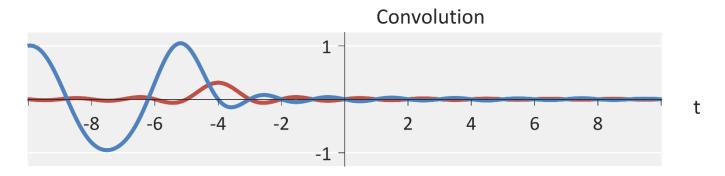


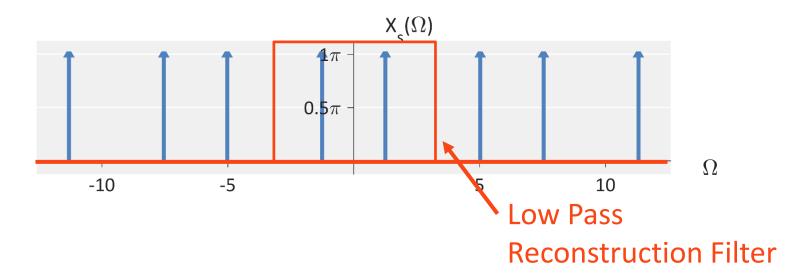


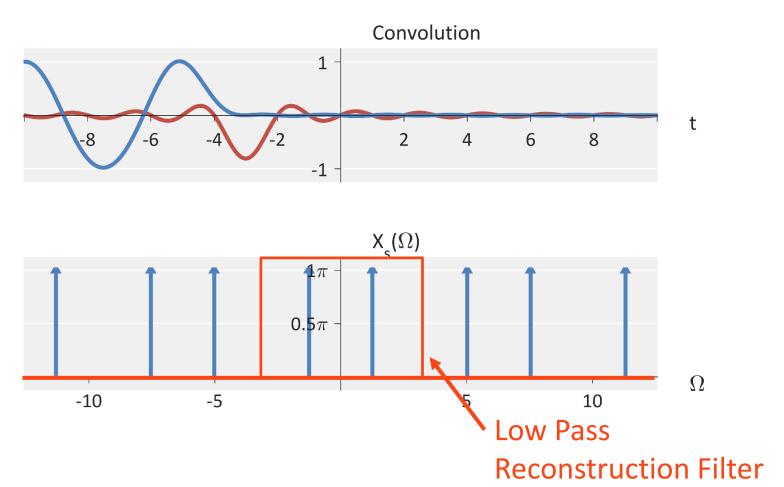


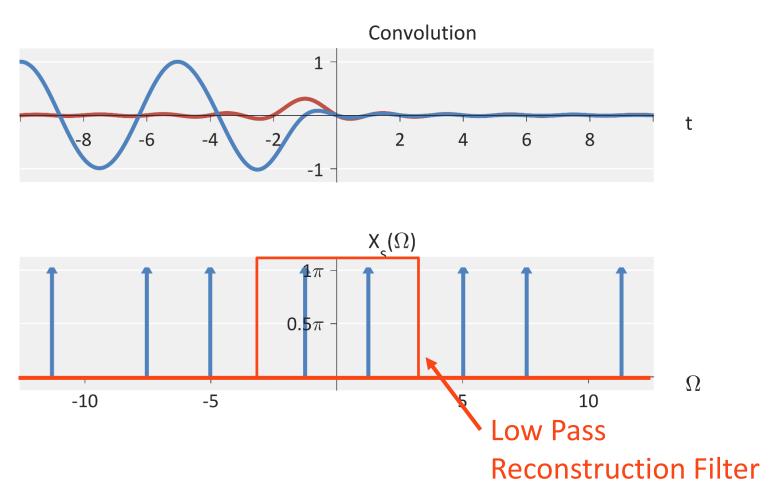


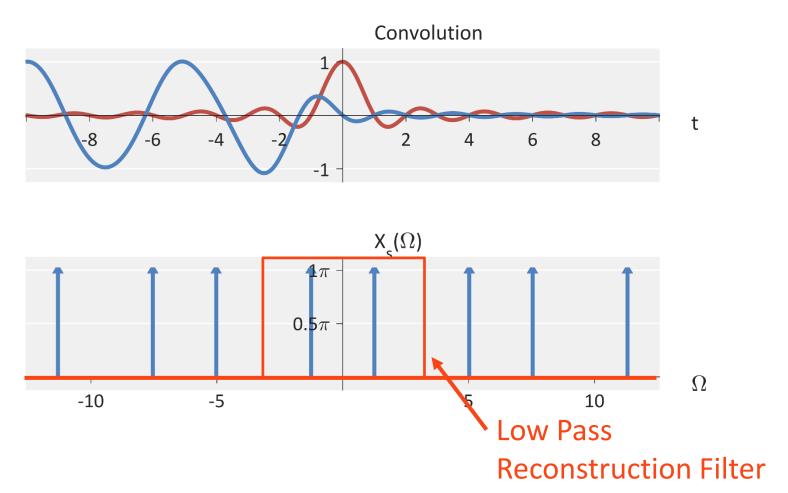


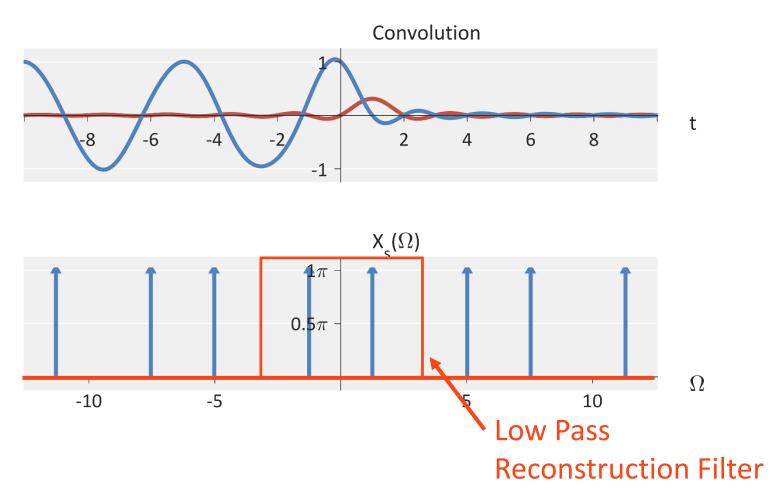


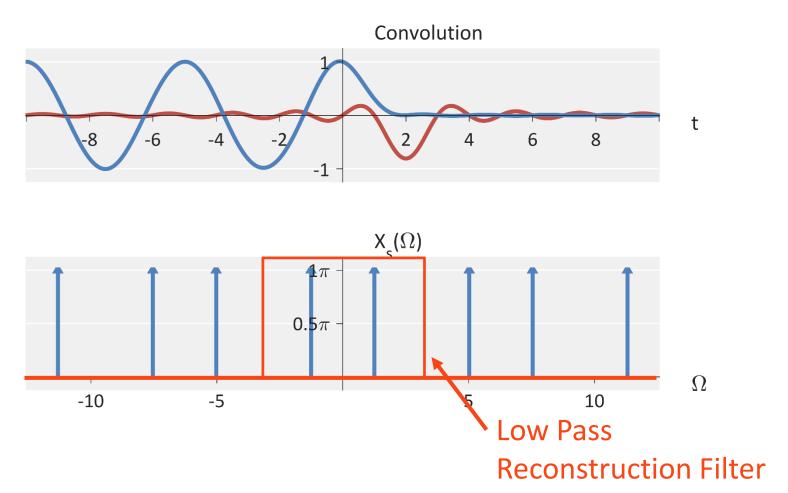


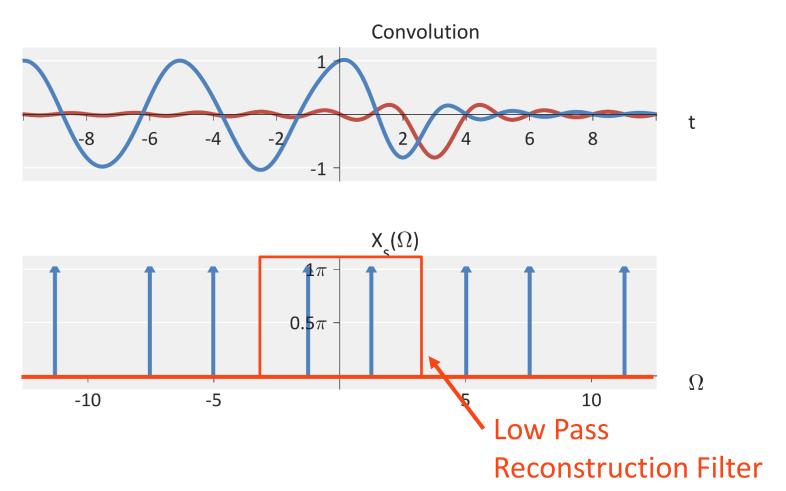


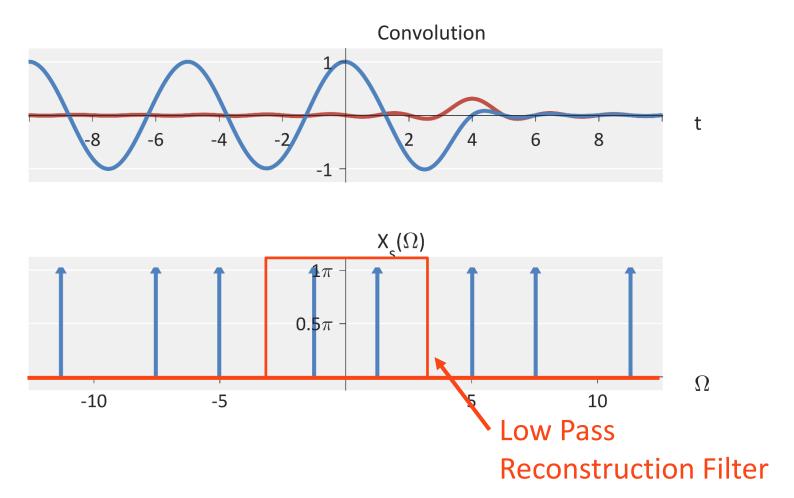


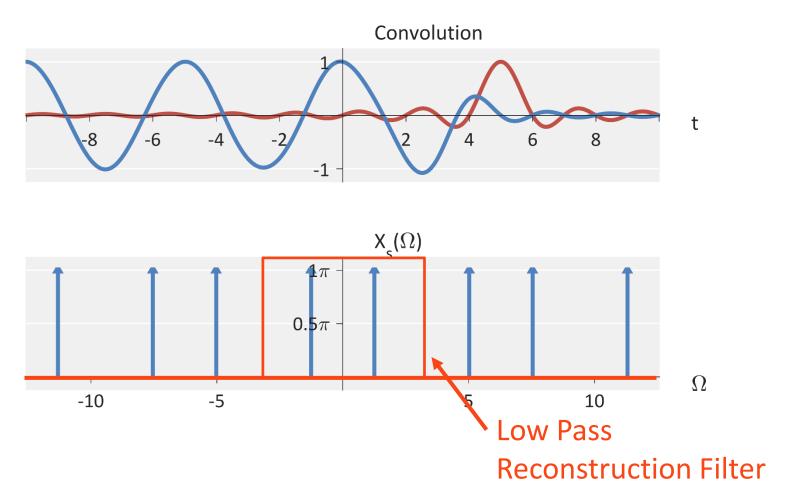


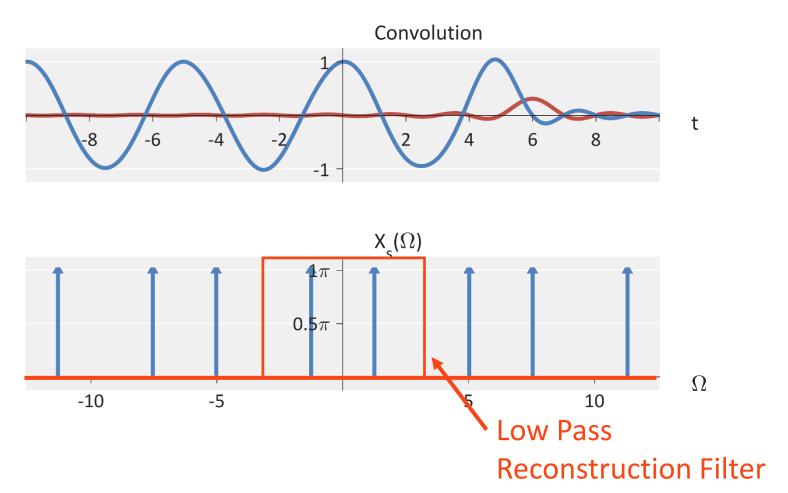


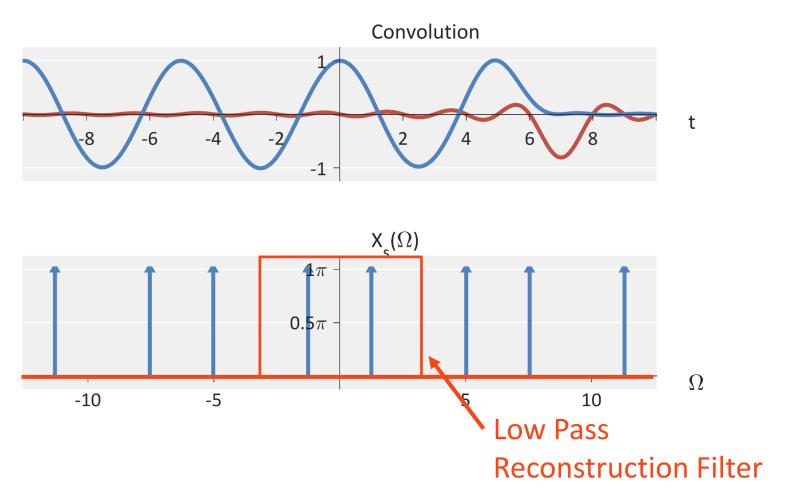


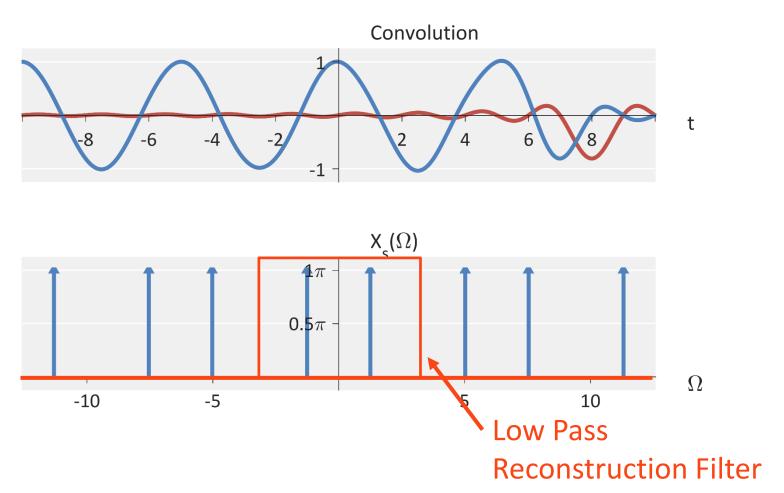


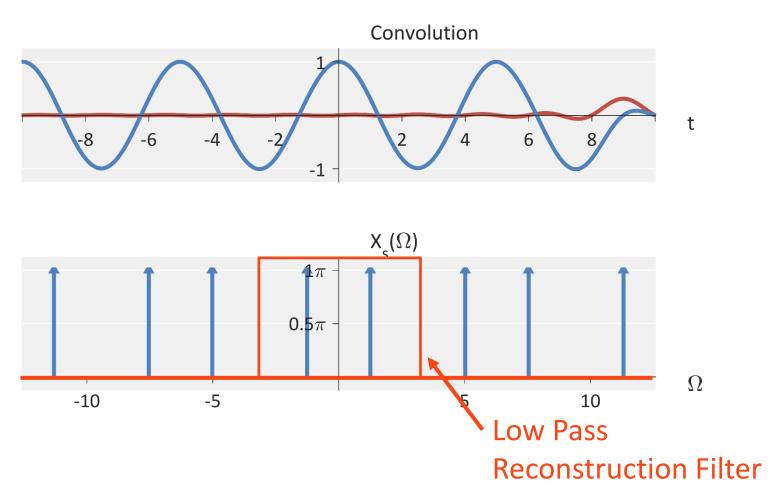


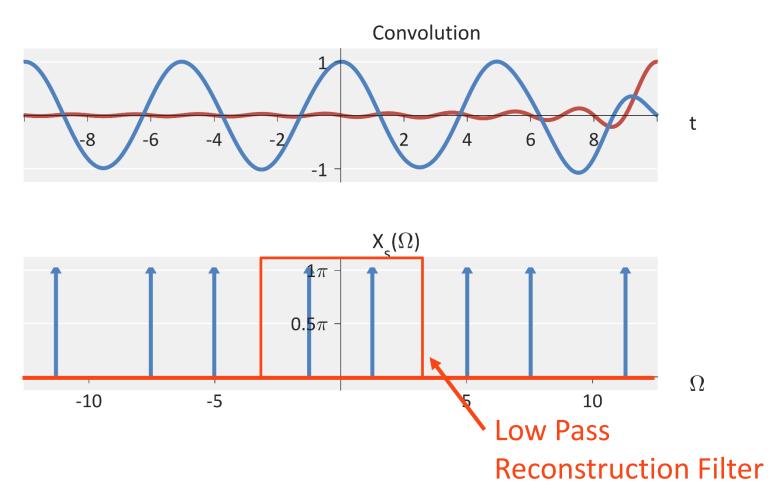


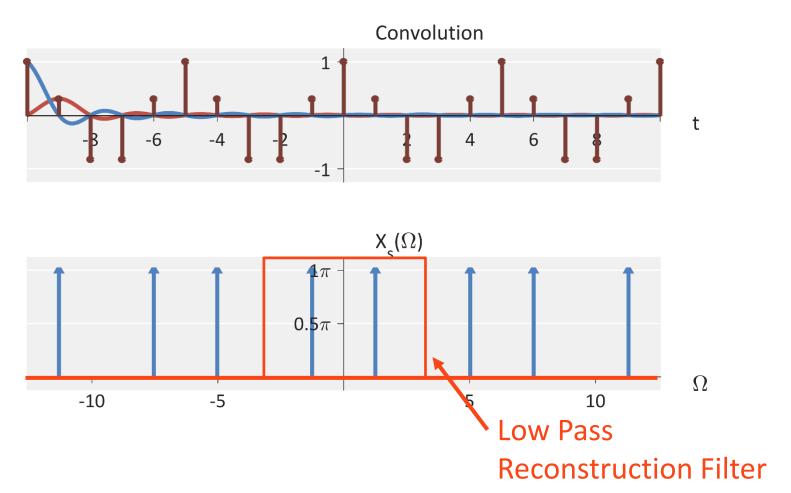


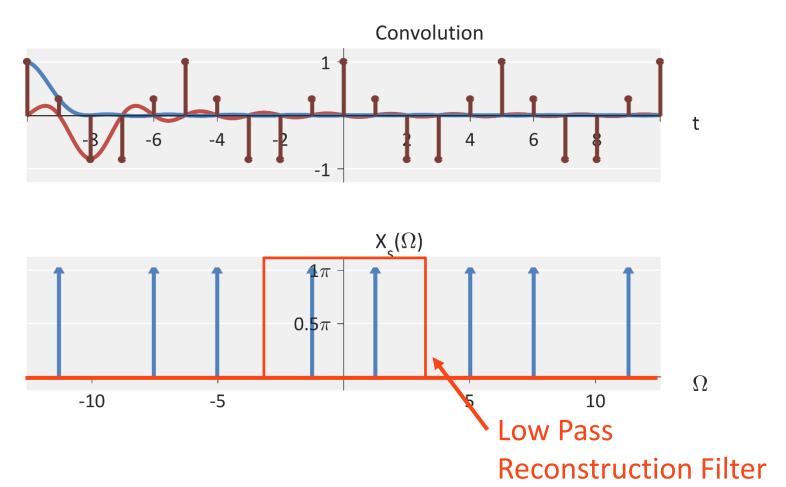


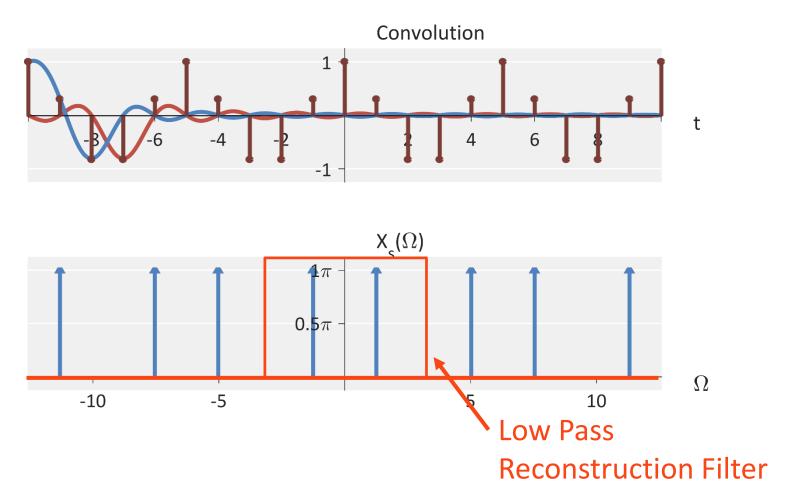


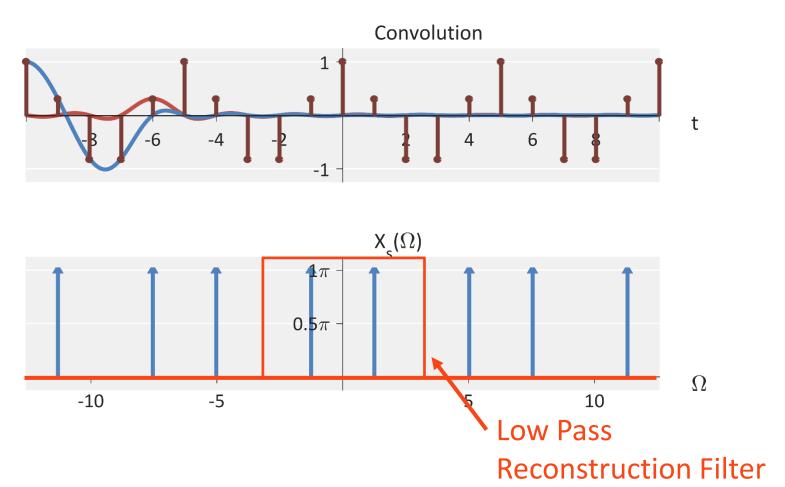


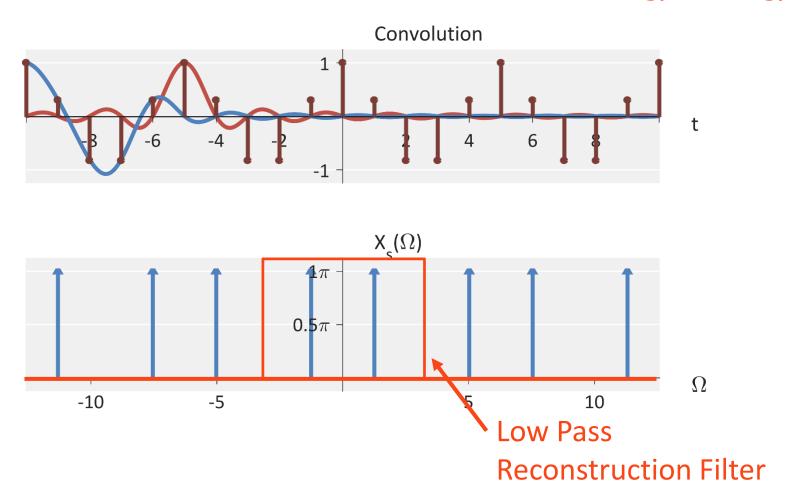


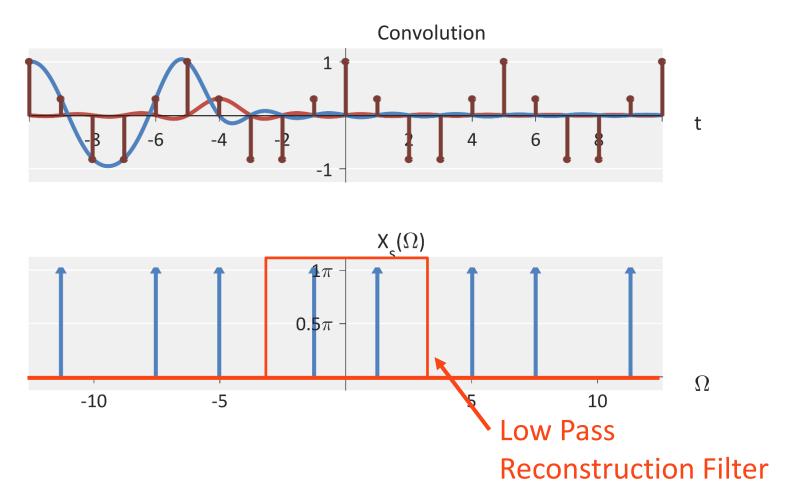


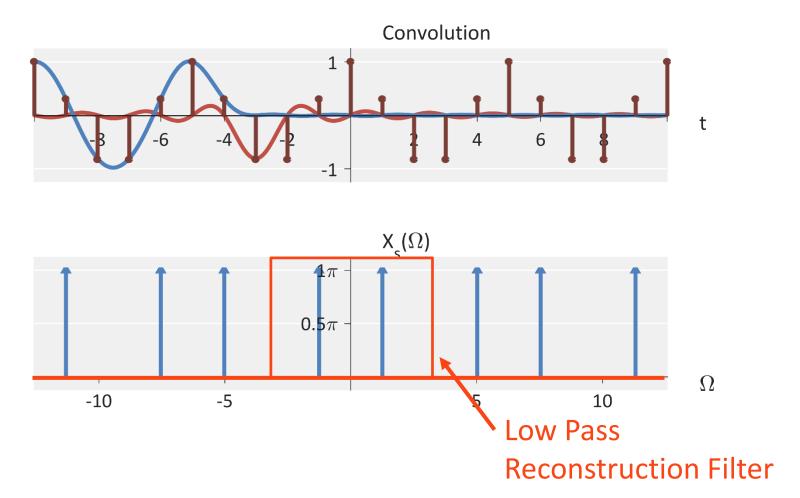


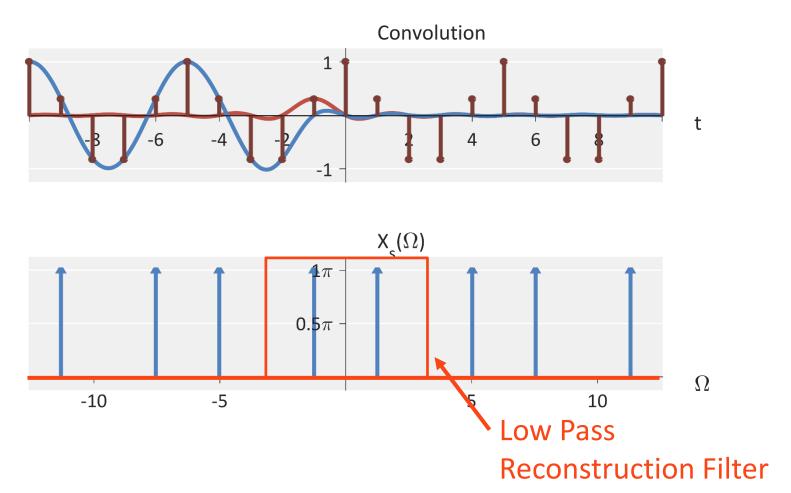


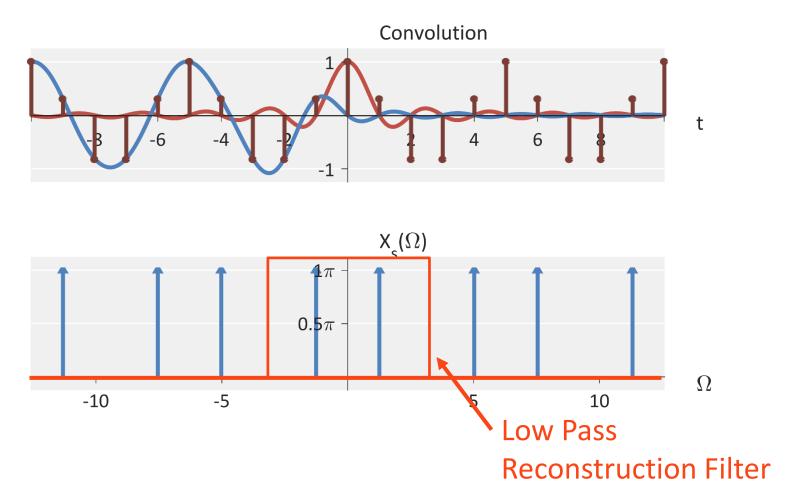


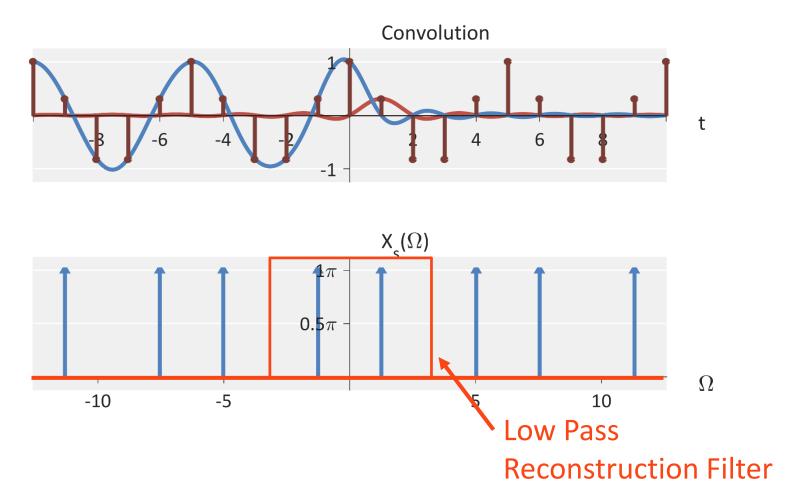


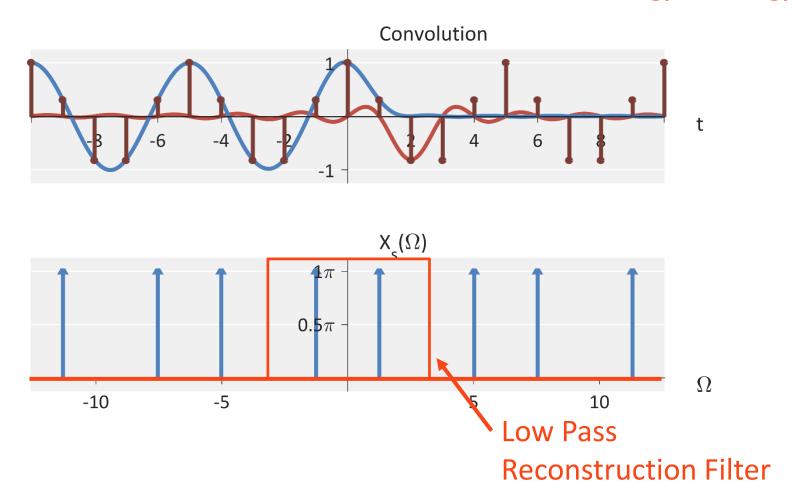


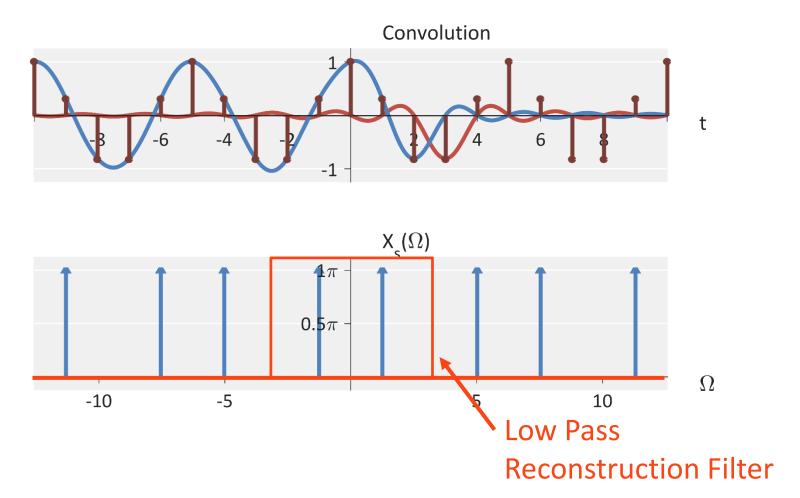


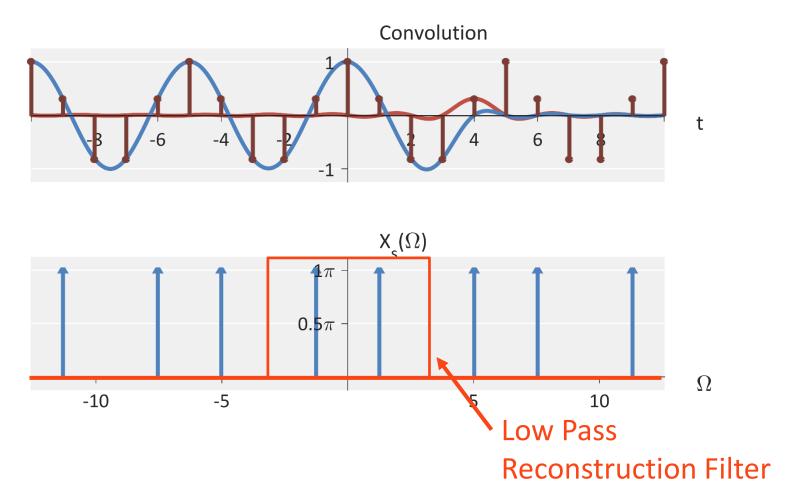


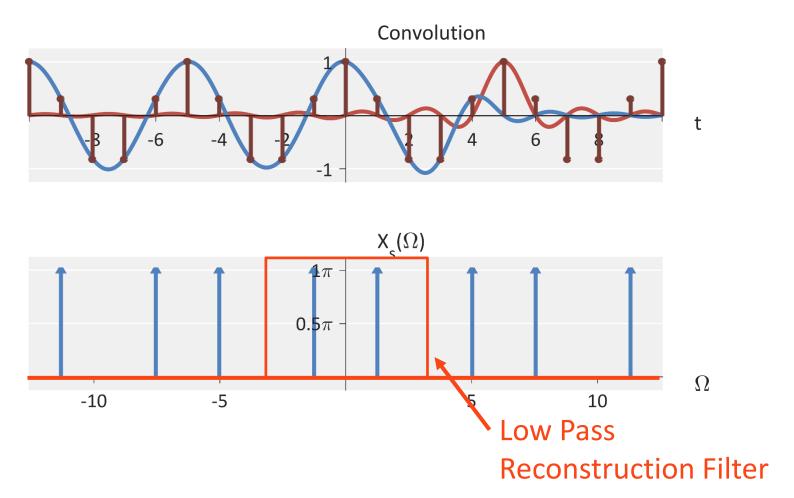


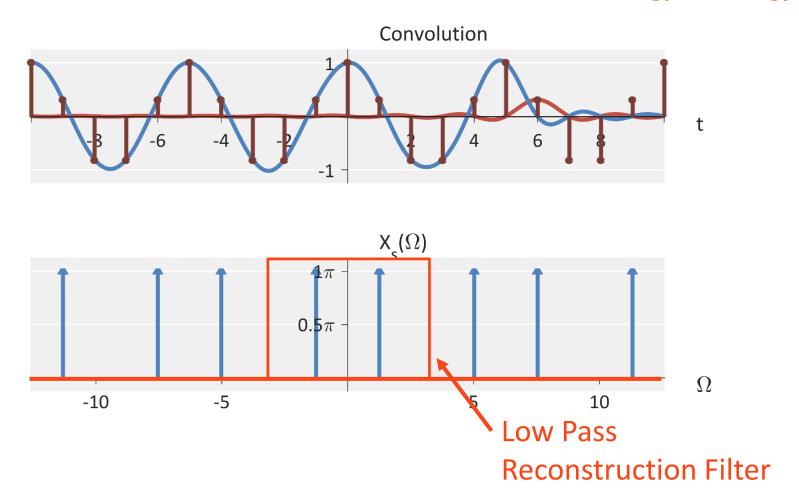


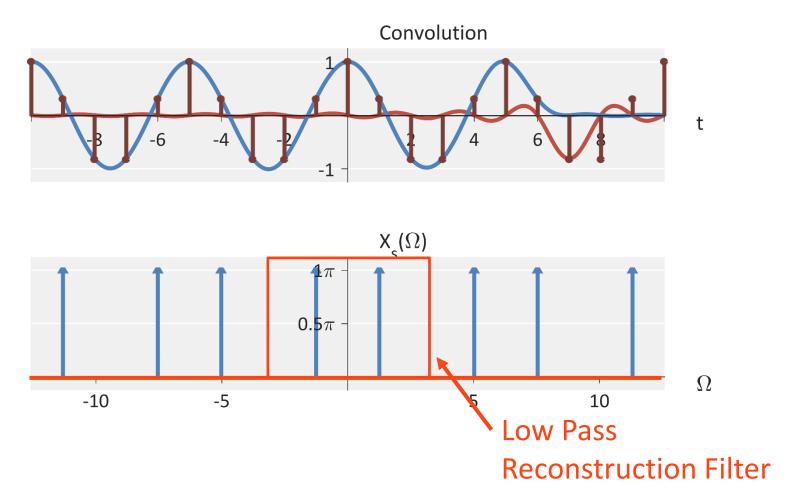


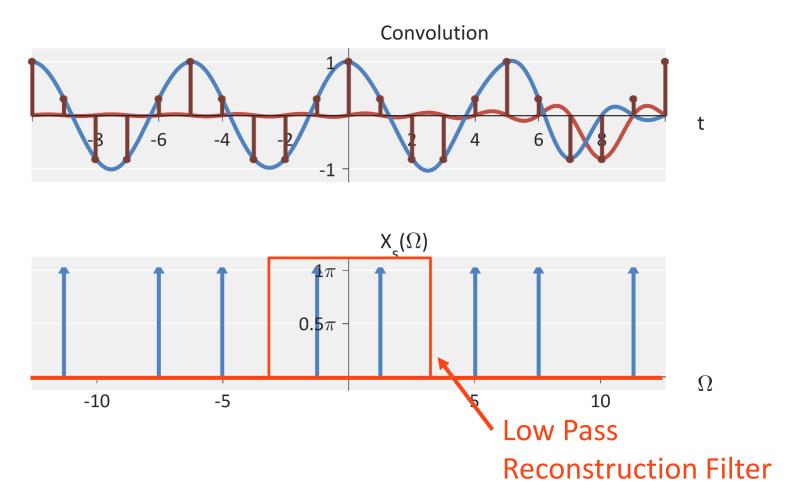


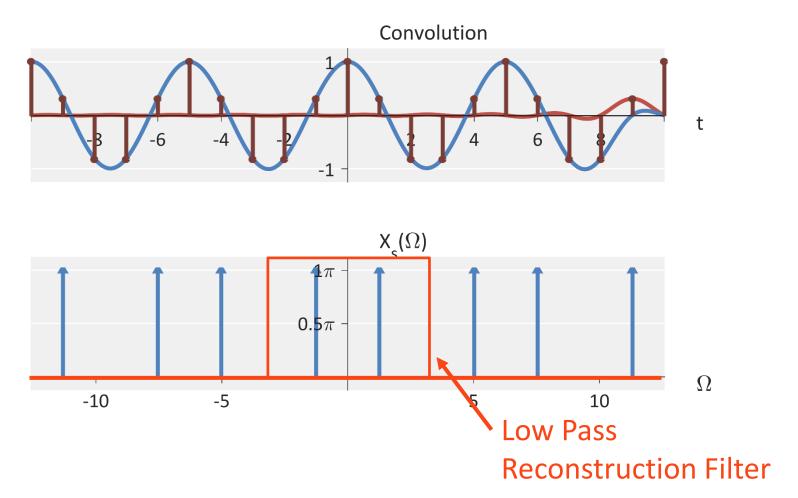


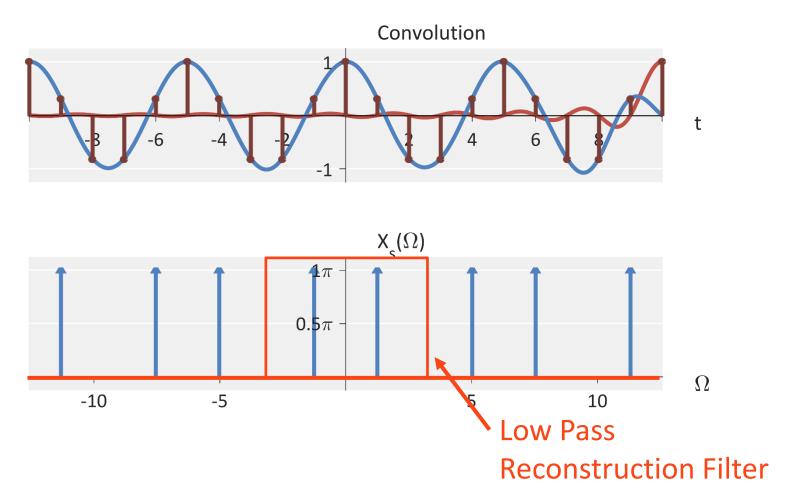




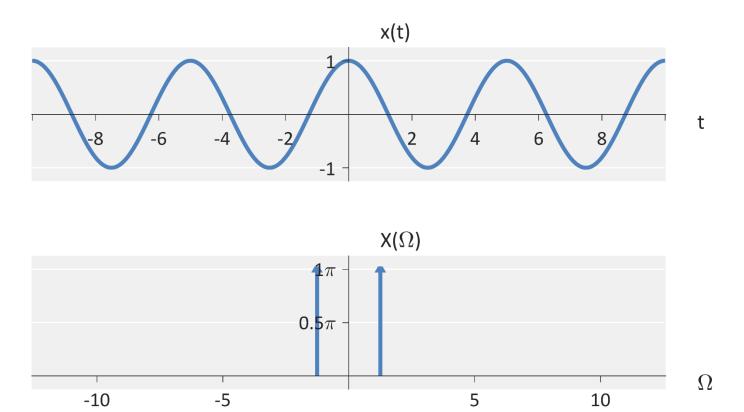








#### ■ Consider a cosine – After the reconstruction filter



# Lecture 12: Sampling, Aliasing, and the Discrete Fourier Transform

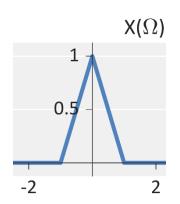
Foundations of Digital Signal Processing

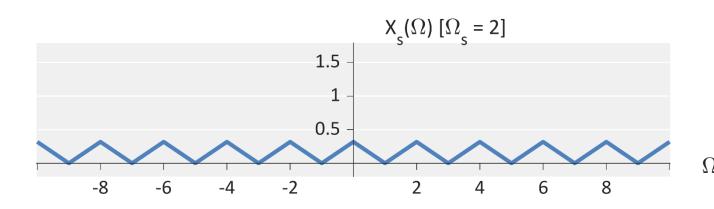
#### **Outline**

- Review of Sampling
- The Nyquist-Shannon Sampling Theorem
- Continuous-time Reconstruction / Interpolation
- Aliasing and anti-Aliasing
- Deriving Transforms from the Fourier Transform
  - Discrete-time Fourier Transform, Fourier Series, Discrete-time Fourier Series
- The Discrete Fourier Transform

# Sampling

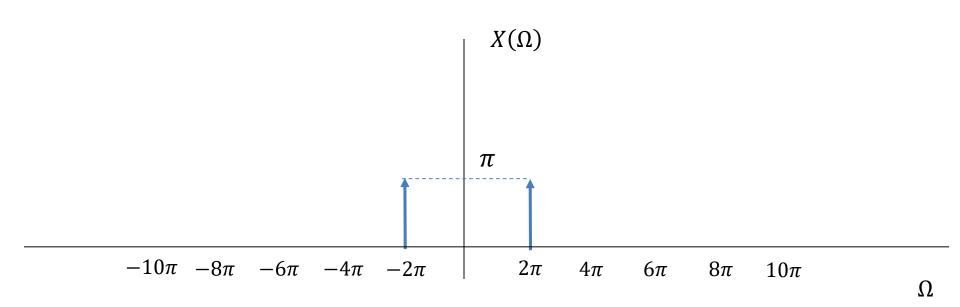
- Aliasing occurs when we do not satisfy the sampling theorem
- Question: What can happen when there is aliasing?



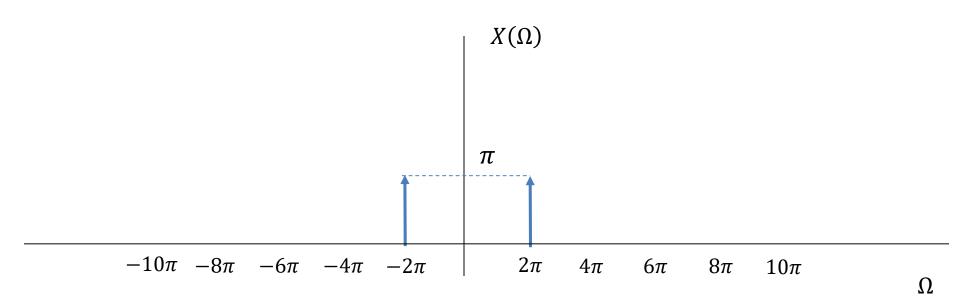


Example 1

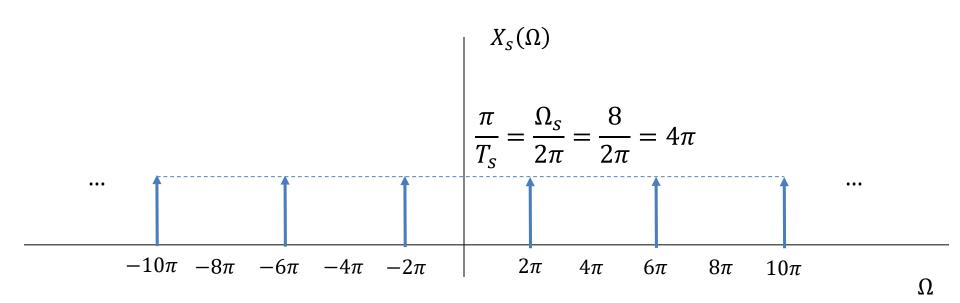
- **Example: Consider**  $x(t) = \cos(2\pi t)$ 
  - Sample this at a rate of  $\Omega_{\scriptscriptstyle S}=8\pi$ 
    - What is the Nyquist rate?
    - What is the cutoff frequency for the low-pass filter?



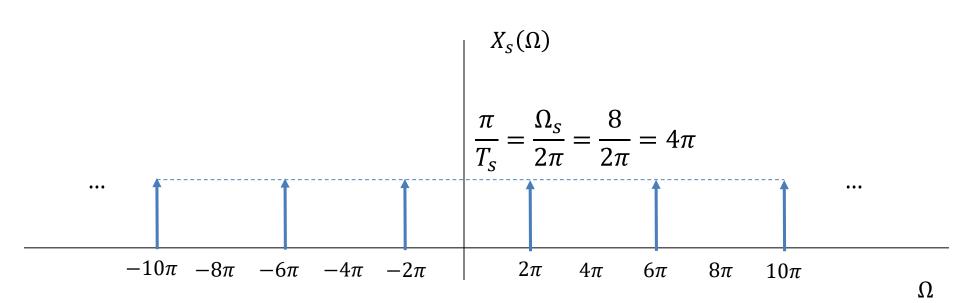
- **Example: Consider**  $x(t) = \cos(2\pi t)$ 
  - Sample this at a rate of  $\Omega_{\scriptscriptstyle S}=8\pi$ 
    - $\diamond$  What is the Nyquist rate?  $\Omega_{\rm S} > 4\pi$
    - $\diamond$  What is the cutoff frequency for the low-pass filter?  $\Omega_c=4\pi$



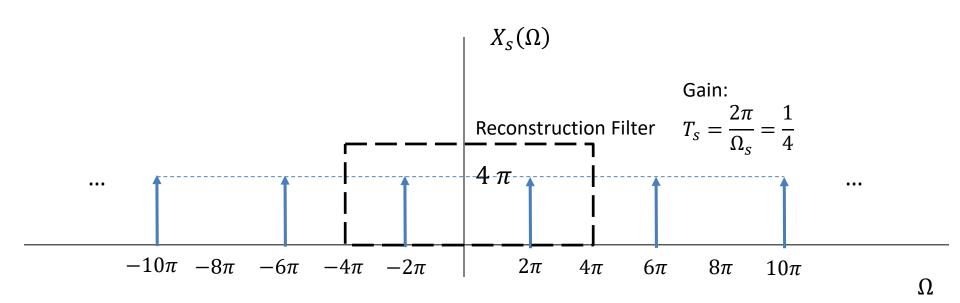
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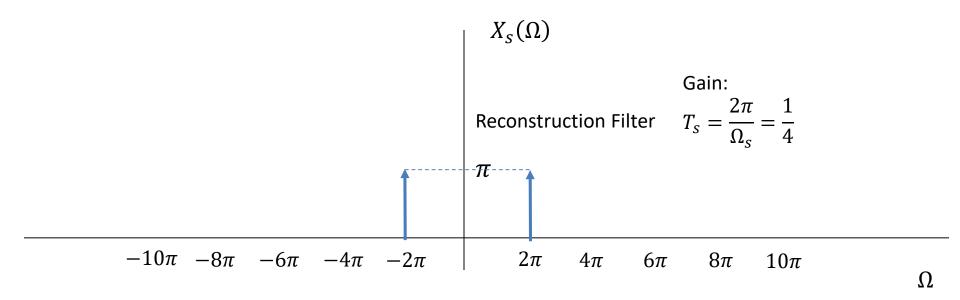
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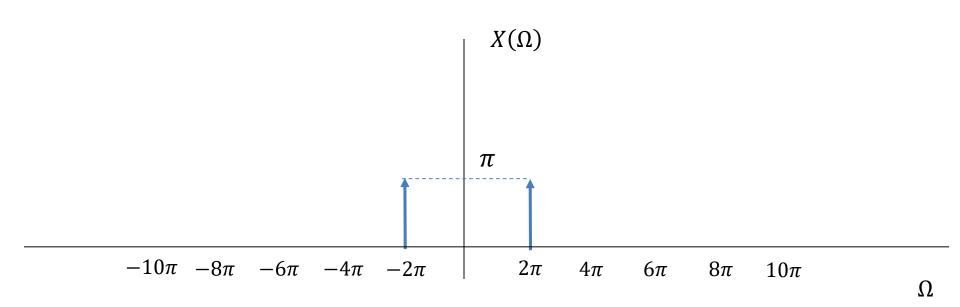


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  - Sample this at a rate of  $\Omega_{\scriptscriptstyle S}=8\pi$ 
    - $\diamond$  What is the Nyquist rate?  $\Omega_{\rm S} > 4\pi$
    - $\diamond$  What is the cutoff frequency for the low-pass filter?  $\Omega_c=4\pi$
    - Reconstructed Signal:  $x(t) = \cos(2\pi t)$

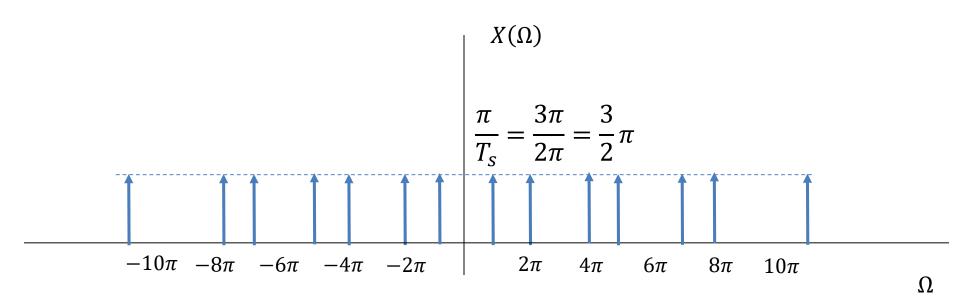


Example 2

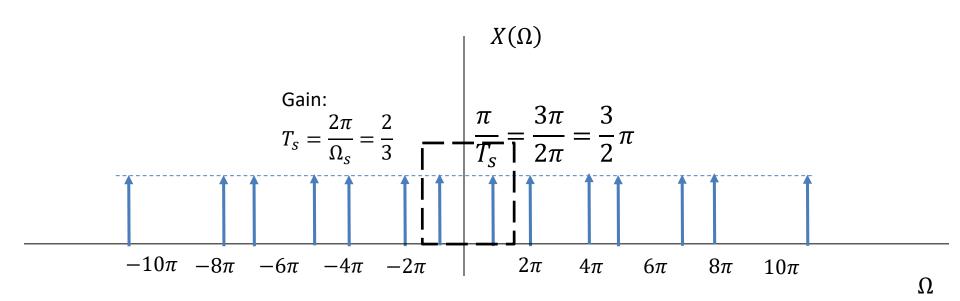
- **Example: Consider**  $x(t) = \cos(2\pi t)$ 
  - Sample this at a rate of  $\Omega_{\scriptscriptstyle S}=3\pi$ 
    - $\diamond$  What is the Nyquist rate?  $\Omega_{\scriptscriptstyle S} > 4\pi$
    - $\diamond$  What is the cutoff frequency for the low-pass filter?  $\Omega_c=1.5\pi$



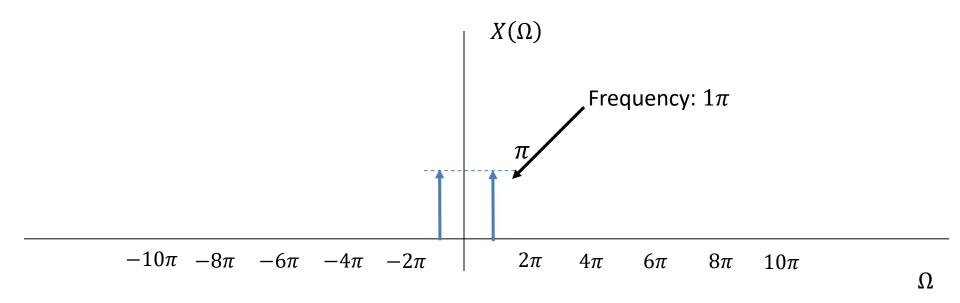
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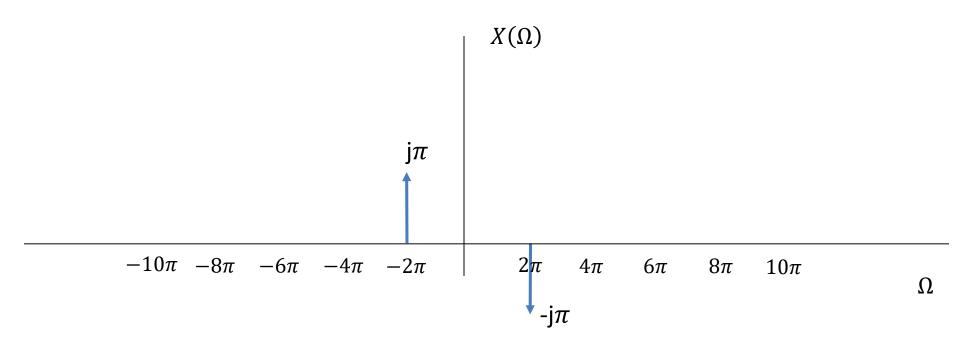


- **Example: Consider**  $x(t) = \cos(2\pi t)$ 
  - Sample this at a rate of  $\Omega_S=3\pi$ 
    - $\diamond$  What is the Nyquist rate?  $\Omega_{S} > 4\pi$
    - $\diamond$  What is the cutoff frequency for the low-pass filter?  $\Omega_c=1.5\pi$
    - Reconstructed Signal:  $x(t) = \cos(\pi t)$

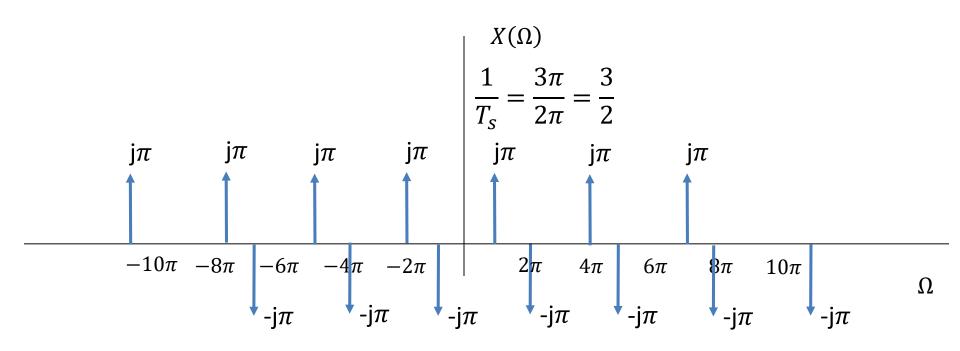


Example 3

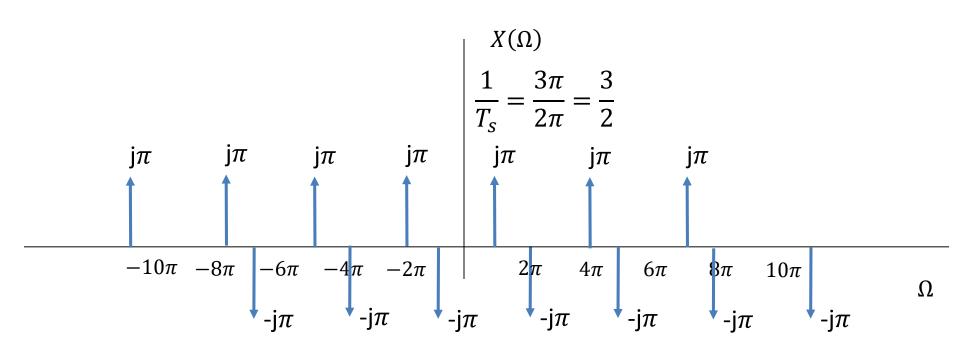
- **Consider**  $x(t) = \sin(2\pi t)$ 
  - Sample this at a rate of  $\Omega_{\scriptscriptstyle S}=3\pi$ 
    - $\diamond$  What is the Nyquist rate?  $\Omega_{S} > 4\pi$
    - $\diamond$  What is the cutoff frequency for the low-pass filter?  $\Omega_c=1.5\pi$



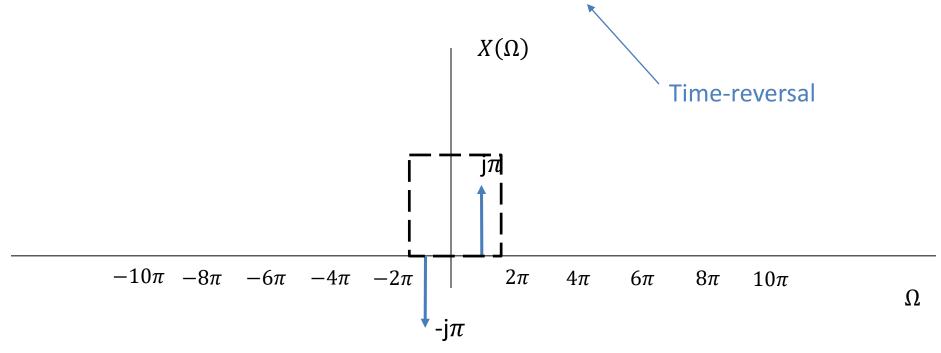
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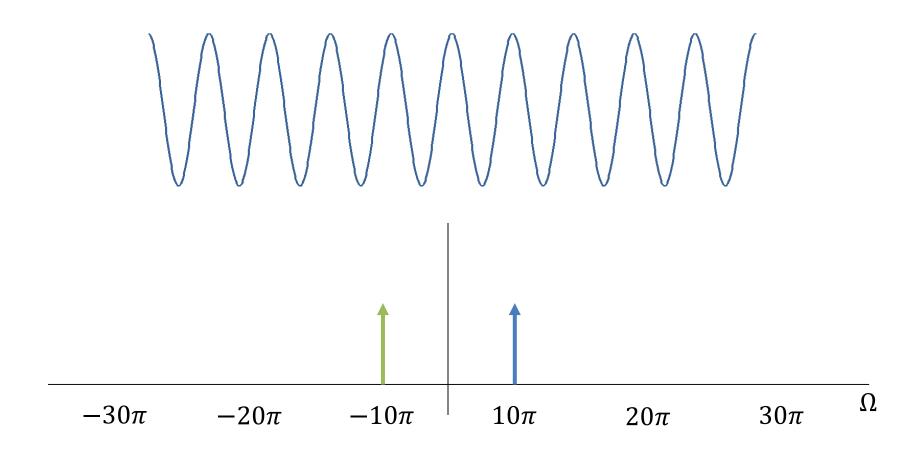


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    - $\diamond$  What is the Nyquist rate?  $\Omega_{S} > 4\pi$
    - $\diamond$  What is the cutoff frequency for the low-pass filter?  $\Omega_c=1.5\pi$
    - Reconstructed Signal:  $x(t) = \sin(-\pi t) = -\sin(\pi t)$

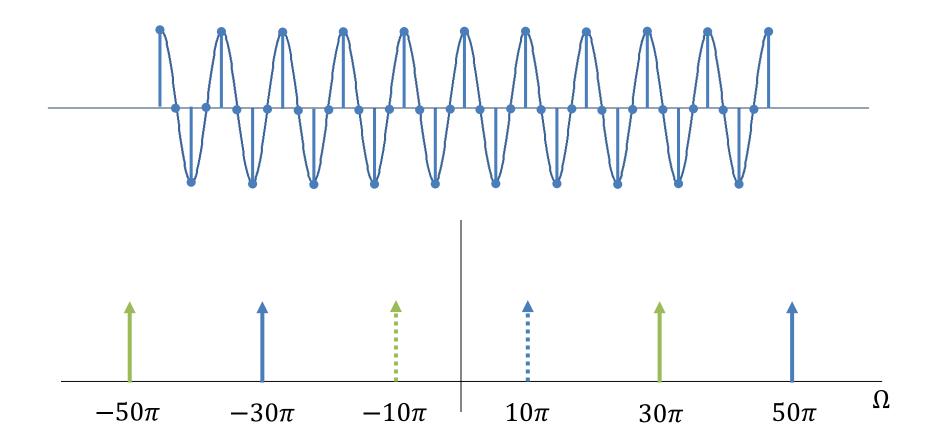


Example 4

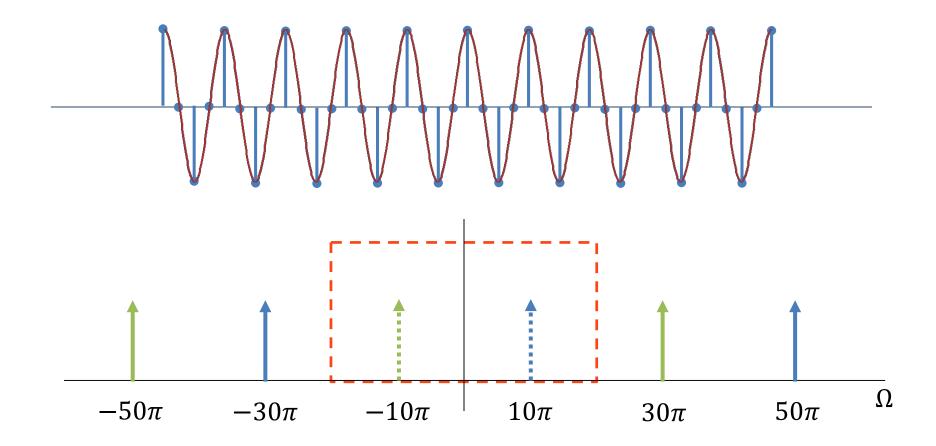
$$x(t) = \cos(10\pi t)$$
 ( $T = 1/5$ )



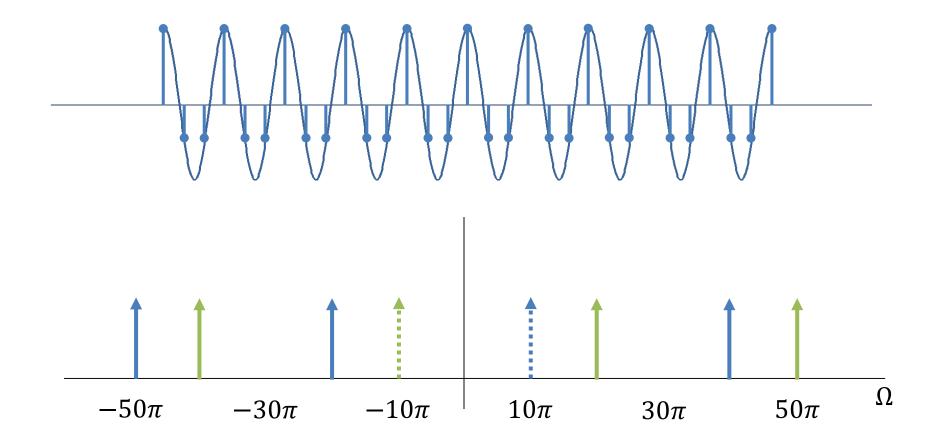
- $x(t) = \cos(10\pi t)$  (T = 1/5)
  - Sample with a sampling rate of  $\Omega_S = 40\pi \ (T_S = 1/20)$



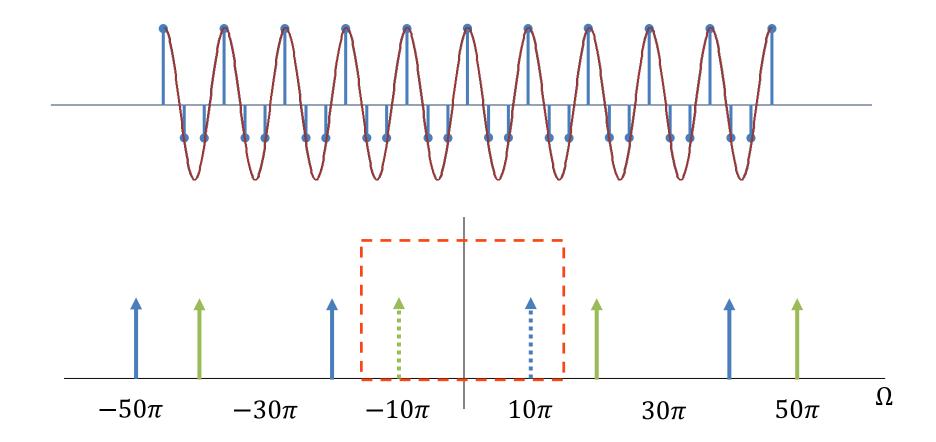
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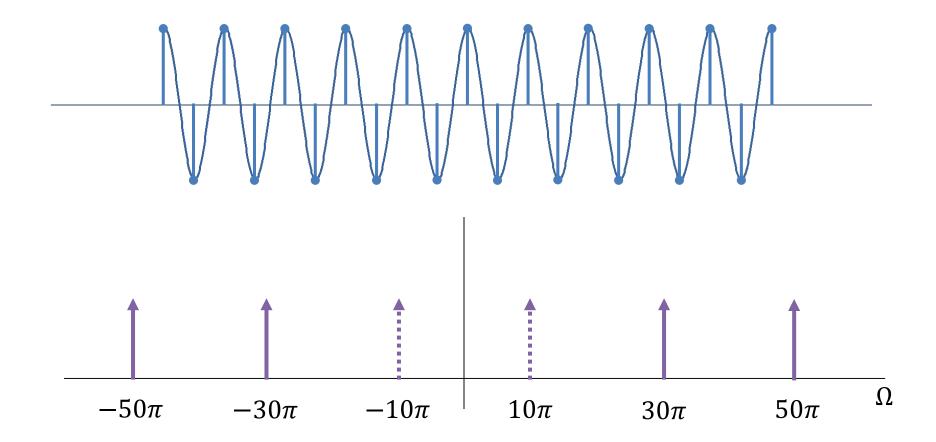
- $x(t) = \cos(10\pi t)$  (T = 1/5)
  - Sample with a sampling rate of  $\Omega_s = 30\pi \ (T_s = 1/15)$



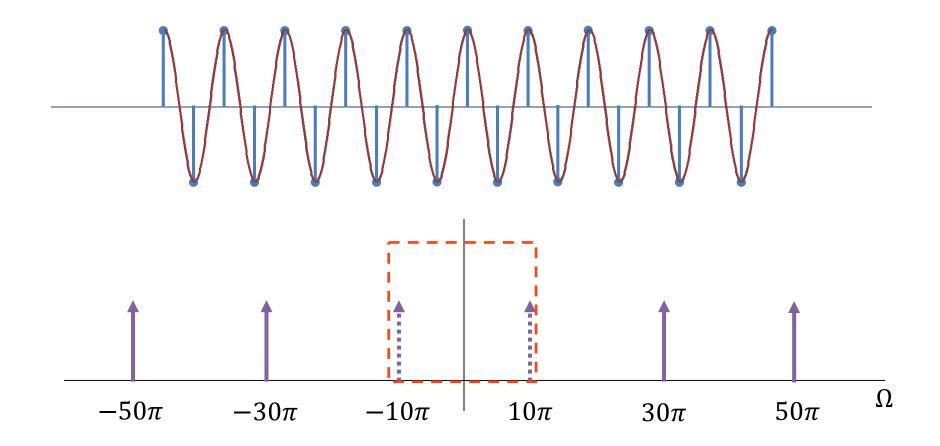
- $x(t) = \cos(10\pi t)$  (T = 1/5)
  - Sample with a sampling rate of  $\Omega_{\rm S}=30\pi~(T_{\rm S}=1/15)$



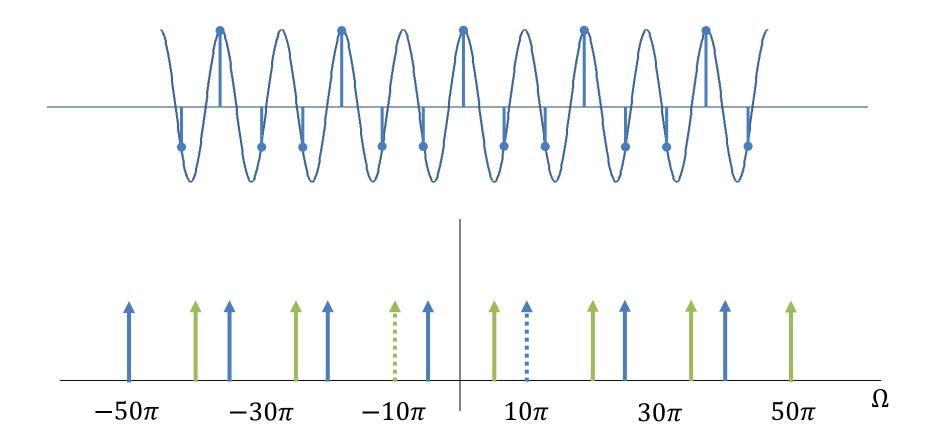
- $x(t) = \cos(10\pi t)$  (T = 1/5)
  - Sample with a sampling rate of  $\Omega_{\scriptscriptstyle S}=20\pi~(T_{\scriptscriptstyle S}=1/10)$



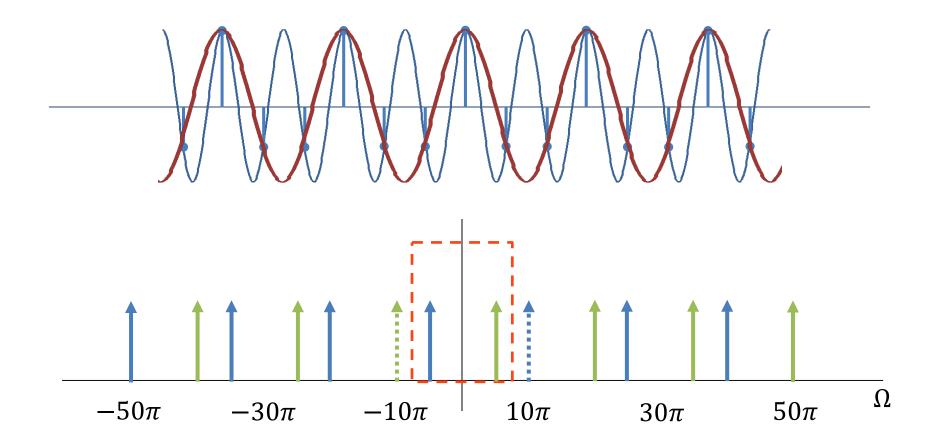
- $x(t) = \cos(10\pi t)$  (T = 1/5)
  - Sample with a sampling rate of  $\Omega_{\scriptscriptstyle S}=20\pi~(T_{\scriptscriptstyle S}=1/10)$



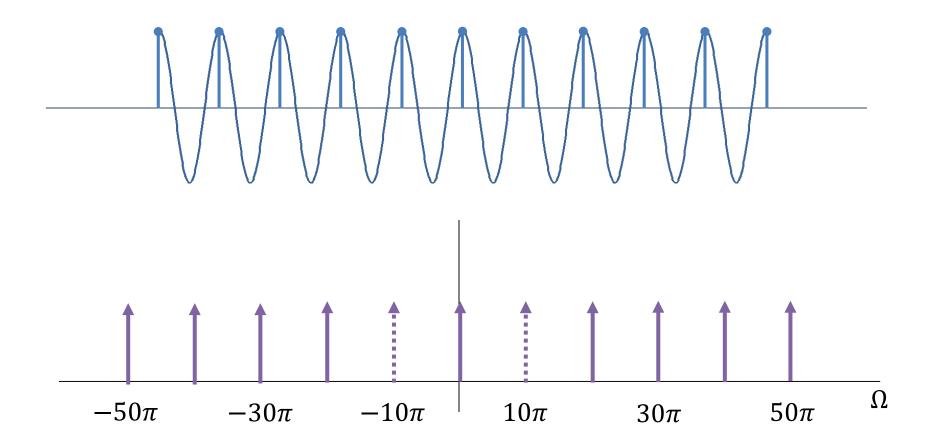
- $x(t) = \cos(10\pi t)$  (T = 1/5)
  - Sample with a sampling rate of  $\Omega_s=15\pi~(T_s=2/15)$



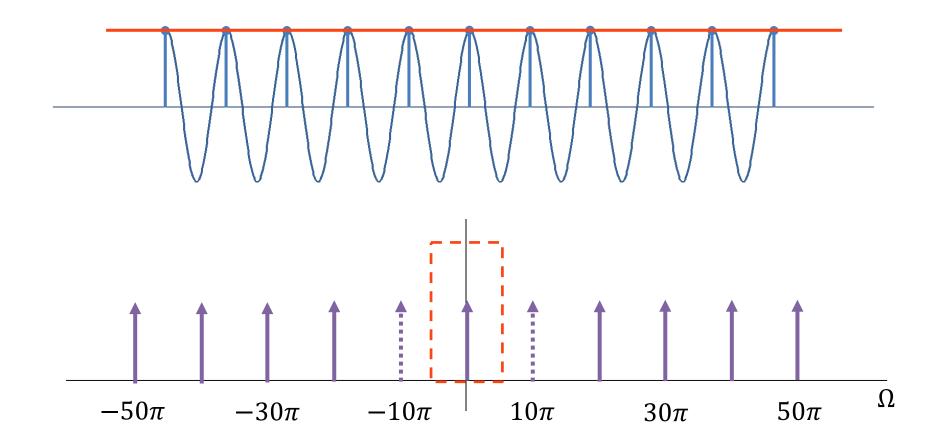
- $x(t) = \cos(10\pi t)$  (T = 1/5)
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- $x(t) = \cos(10\pi t)$  (T = 1/5)
  - Sample with a sampling rate of  $\Omega_{\scriptscriptstyle S}=10\pi \ (T_{\scriptscriptstyle S}=1/5)$

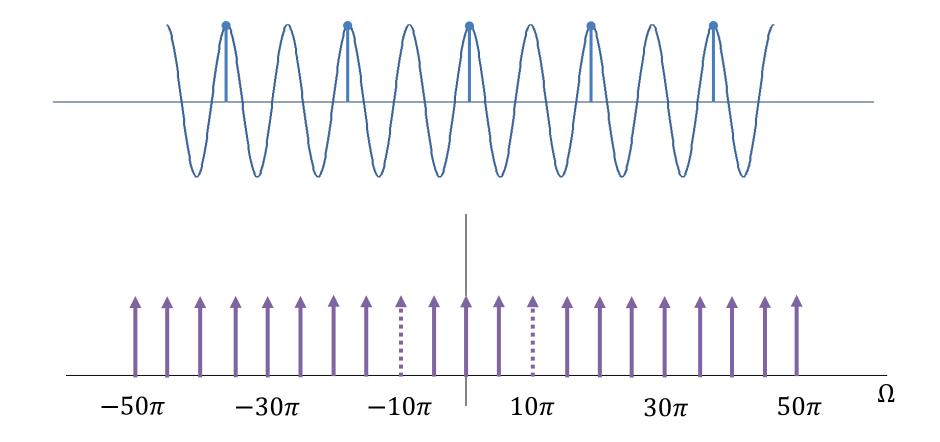


- $x(t) = \cos(10\pi t)$  (T = 1/5)
  - Sample with a sampling rate of  $\Omega_{\rm S}=10\pi~(T_{\rm S}=1/5)$



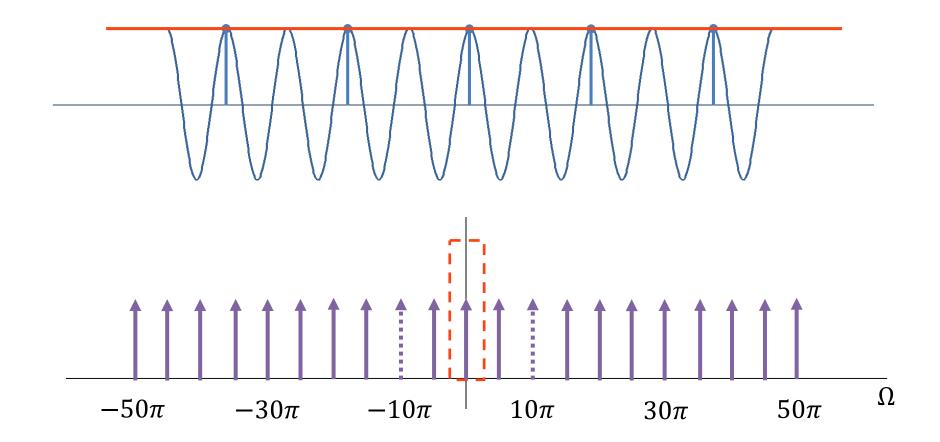
### Aliasing with a sinusoid

- $x(t) = \cos(10\pi t)$  (T = 1/5)
  - Sample with a sampling rate of  $\Omega_{\rm S}=5\pi$   $(T_{\rm S}=2/5)$



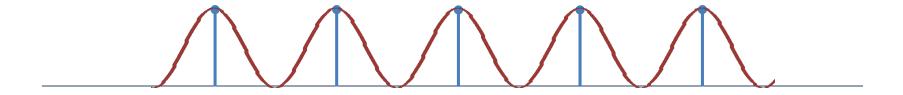
### Aliasing with a sinusoid

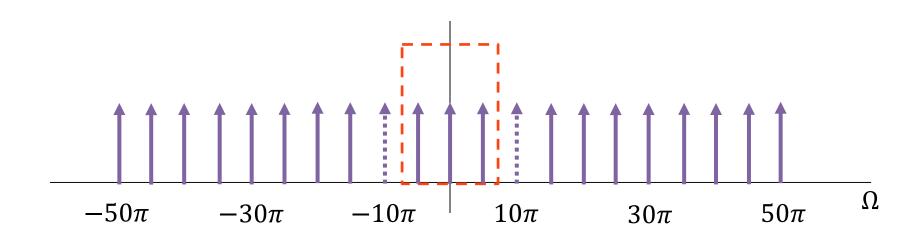
- $x(t) = \cos(10\pi t)$  (T = 1/5)
  - Sample with a sampling rate of  $\Omega_{\rm S}=5\pi$   $(T_{\rm S}=2/5)$



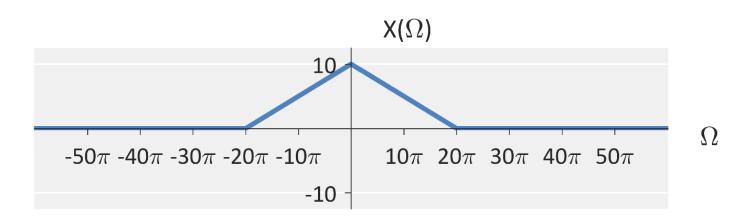
### Aliasing with a sinusoid

- $x(t) = \cos(10\pi t)$  (T = 1/5)
  - Sample with a sampling rate of  $\Omega_{\rm S}=5\pi~(T_{\rm S}=2/5)$

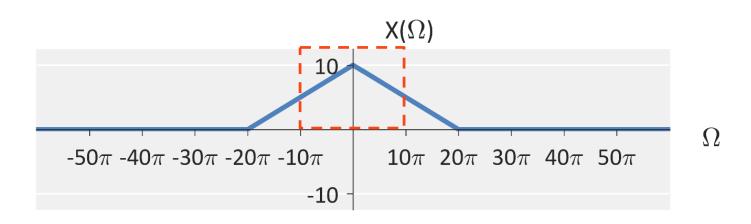




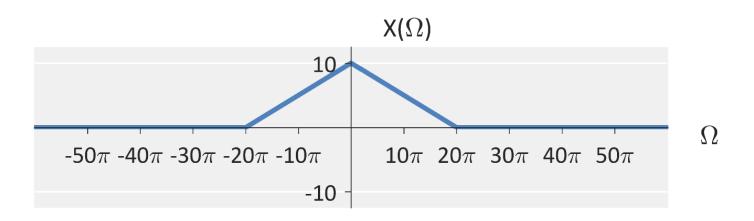
Question: How do we reduce the effects of aliasing?



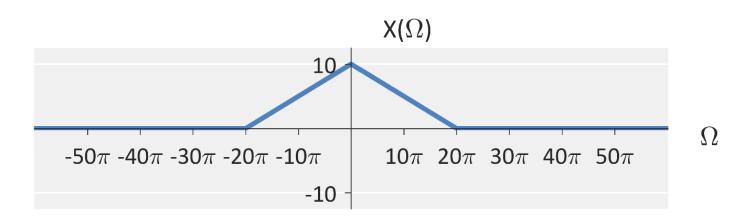
- Question: How do we reduce the effects of aliasing?
  - Apply a low-pass anti-aliasing filter
    - Cut-off frequency:  $\Omega_s/2$
    - Gain: 1



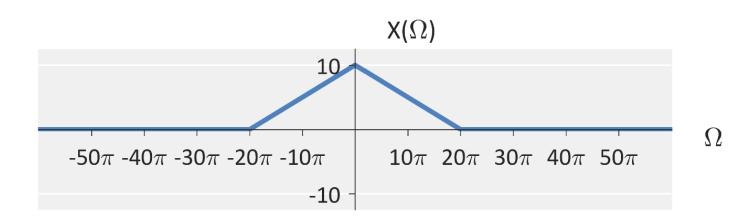
- Example: Consider the following signal.
  - What is the Nyquist rate?



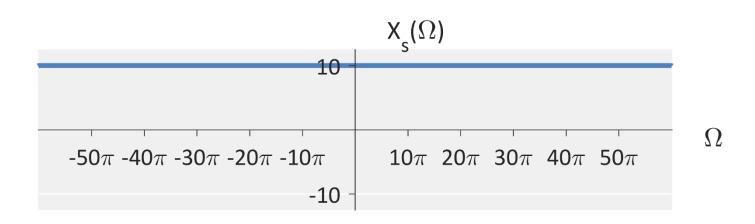
- Example: Consider the following signal.
  - What is the Nyquist rate?  $40\pi$



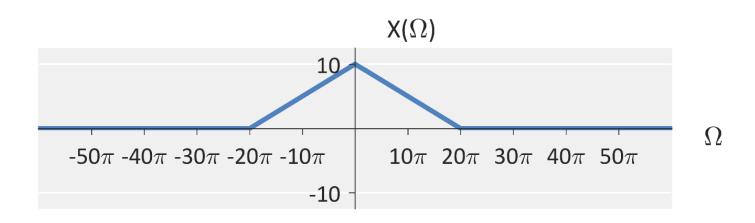
- **Example:** Consider the following signal.
  - What is the Nyquist rate?  $40\pi$
  - Sketch the Fourier transform after sampling at  $\Omega_s=20\pi$ .
  - Use no anti-aliasing filter



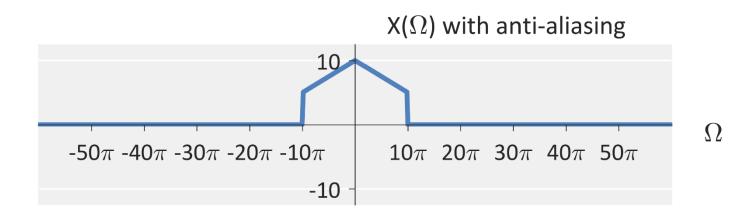
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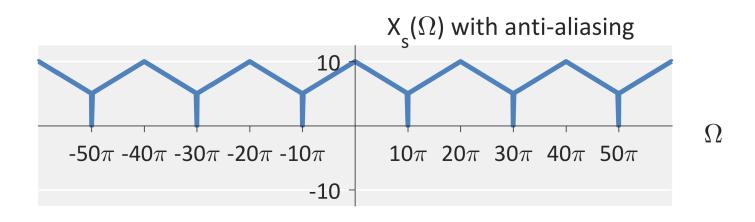
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- **Example:** Consider the following signal.
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  - Use an anti-aliasing filter



Sampling and Aliasing in Real Life

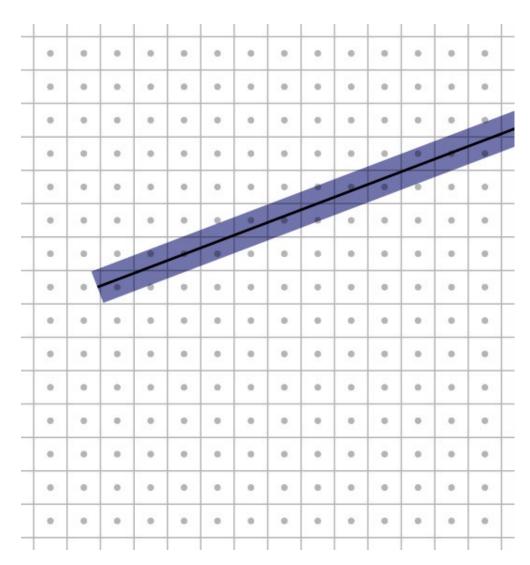
# Aliasing

■ Where have we seen such effects before in real life?

### Sampling

### Sampling a line

We sample a continuous image at each point



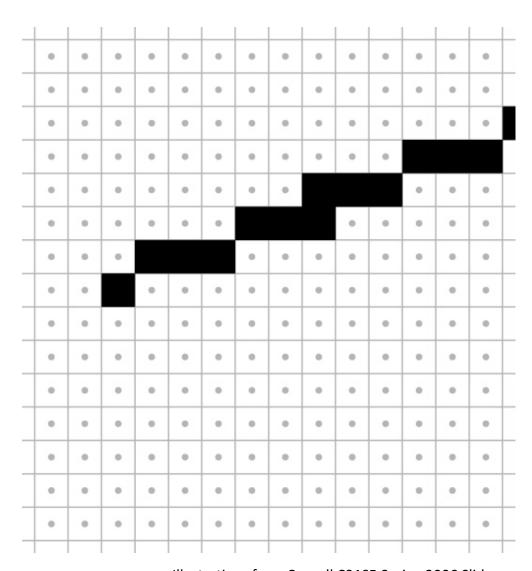
### Sampling

### Sampling a line

We sample a continuous image at each point

### Aliasing

Creates undesired high frequency information



### Sampling a line

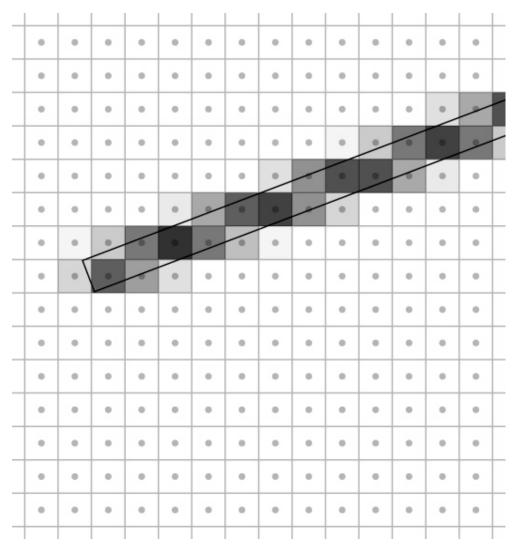
We sample a continuous image at each point

### Aliasing

Creates undesired high frequency information

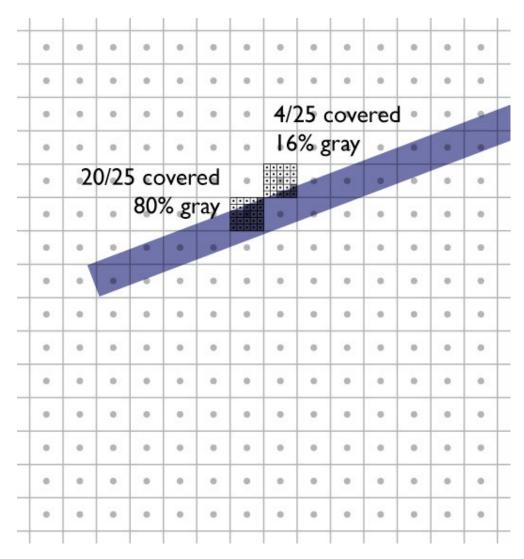
#### Anti-aliasing

Create a smooth image



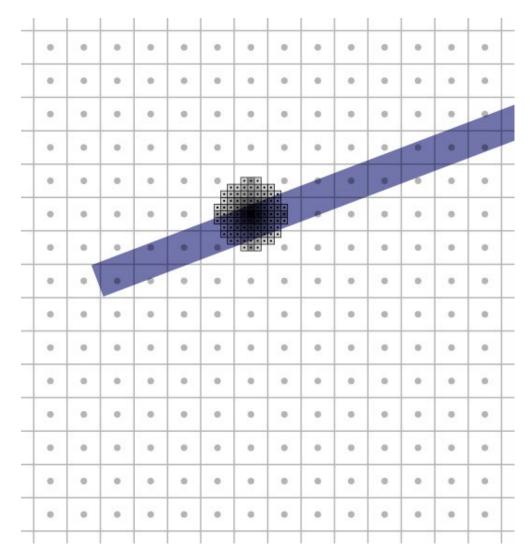
#### Box Filter

 Convolve the continuous (high resolution) image with a box

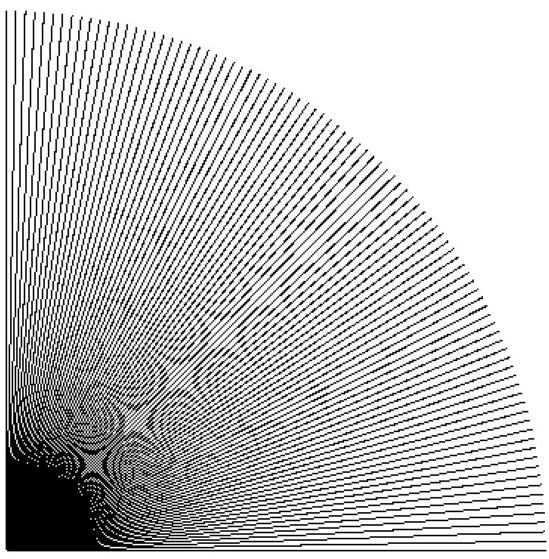


#### Gaussian Filter

 Convolve the continuous (high resolution) image with a Gaussian

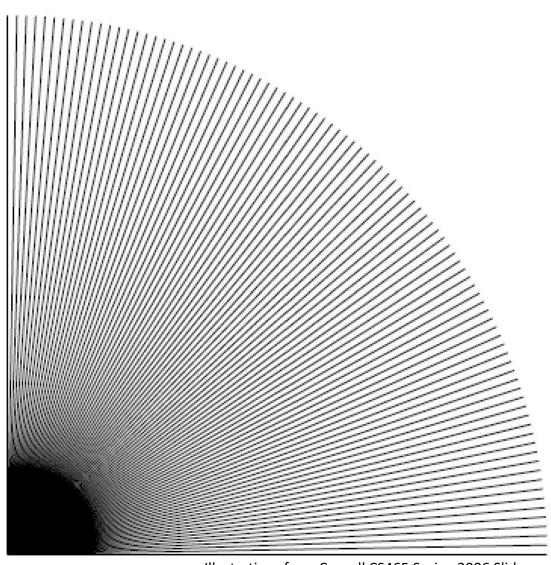


Sampling of lines



### Sampling of lines

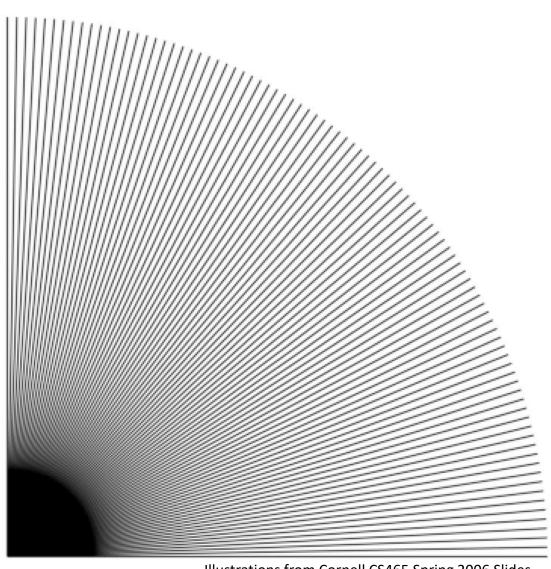
Anti-aliasing with a box filter



Illustrations from Cornell CS465 Spring 2006 Slides

### Sampling of lines

Anti-aliasing with a Gaussian filter



### Aliasing Example

#### A Wheel



From: https://www.youtube.com/watch?v=bl8lrqBBAXQ

# Aliasing Example

### Making water float



From: https://www.youtube.com/watch?v=mODqQvlrgIQ

# Lecture 12: Sampling, Aliasing, and the Discrete Fourier Transform

Foundations of Digital Signal Processing

#### **Outline**

- Review of Sampling
- The Nyquist-Shannon Sampling Theorem
- Continuous-time Reconstruction / Interpolation
- Aliasing and anti-Aliasing
- Deriving Transforms from the Fourier Transform
  - Discrete-time Fourier Transform, Fourier Series, Discrete-time Fourier Series
- The Discrete Fourier Transform

- Consider the Fourier Transform....
  - What happens if we sample x(t)?

$$X(\Omega) = \int_{-\infty}^{\infty} x(t)e^{-j\Omega t} dt$$

- Consider the Fourier Transform....
  - What happens if we sample x(t)?

$$X(\Omega) = \int_{-\infty}^{\infty} x(t) \delta_{T_s}(t) e^{-j\Omega t} dt$$

#### Consider the Fourier Transform....

• What happens if we sample x(t)?

$$X(\Omega) = \int_{-\infty}^{\infty} x(t)\delta_{T_S}(t)e^{-j\Omega t} dt$$

$$= \int_{-\infty}^{\infty} x(t) \left[ \sum_{n=-\infty}^{\infty} \delta(t - nT_S) \right] e^{-j\Omega t} dt$$

$$= \sum_{n=-\infty}^{\infty} \int_{-\infty}^{\infty} x(t)\delta(t - nT_S)e^{-j\Omega t} dt$$

#### Consider the Fourier Transform....

• What happens if we sample x(t)?

$$X(\Omega) = \int_{-\infty}^{\infty} x(t)\delta_{T_s}(t)e^{-j\Omega t} dt$$

$$= \int_{-\infty}^{\infty} x(t) \left[ \sum_{n=-\infty}^{\infty} \delta(t - nT_s) \right] e^{-j\Omega t} dt$$

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

#### Consider the Fourier Transform....

• What happens if we sample x(t)?

$$X(\Omega) = \int_{-\infty}^{\infty} x(t)\delta_{T_S}(t)e^{-j\Omega t} dt$$

$$= \int_{-\infty}^{\infty} x(t) \left[ \sum_{n=-\infty}^{\infty} \delta(t - nT_S) \right] e^{-j\Omega t} dt$$

$$= \sum_{n=-\infty}^{\infty} x(nT_S)e^{-j\Omega nT_S} \quad \text{Choose } T_S = 1, \Omega = \omega$$

- Consider the Fourier Transform....
  - What happens if we sample x(t)?

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

The Fourier Transform becomes the DTFT

- Consider the Inverse Fourier Transform....
  - What happens if we sample  $X(\Omega)$ ?

$$x(t) = \int_{-\infty}^{\infty} X(\Omega) e^{+j\Omega t} d\Omega$$

#### Consider the Inverse Fourier Transform....

• What happens if we sample  $X(\Omega)$ ?

$$x(t) = \int_{-\infty}^{\infty} X(\Omega)e^{+j\Omega t} d\Omega$$

$$= \int_{-\infty}^{\infty} X(\Omega) \left[ \sum_{k=-\infty}^{\infty} \delta(\Omega - k\Omega_s) \right] e^{+j\Omega t} d\Omega$$

$$= \sum_{k=-\infty}^{\infty} X(k\Omega_s)e^{+jk\Omega_s t} \qquad \text{Choose } X(k\Omega_s) = c_k$$

- Consider the Inverse Fourier Transform....
  - What happens if we sample  $X(\Omega)$ ?

$$x(t) = \sum_{k=-\infty}^{\infty} c_k e^{+jk\Omega_S t}$$

The Fourier Transform becomes the Fourier Series

- Consider the Inverse Discrete-Time Fourier Transform....
  - What happens if we sample  $X(\omega)$ ?

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

- Consider the Inverse Discrete-Time Fourier Transform....
  - What happens if we sample  $X(\omega)$ ?

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

$$= \frac{1}{2\pi} \int_{2\pi} X(\omega) \left[ 2\pi \sum_{k=-\infty}^{\infty} \delta(\omega - k\omega_s) \right] e^{+j\omega n} d\omega$$

- Consider the Inverse Discrete-Time Fourier Transform....
  - What happens if we sample  $X(\omega)$ ?

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

$$= \frac{1}{2\pi} \int_{0}^{2\pi} X(\omega) \left[ 2\pi \sum_{k=-\infty}^{\infty} \delta(\omega - k\omega_s) \right] e^{+j\omega n} d\omega$$

$$k\omega_s \ge 0 \qquad k\omega_s < 2\pi$$

$$k \ge 0 \qquad k < \frac{2\pi}{\omega_s}$$

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$$= \sum_{k=0}^{\frac{2\pi}{\omega_s} - 1} X(k\omega) e^{+jk\omega_s n}$$

$$= \sum_{k=0}^{\infty} X(k\omega) e^{+jk\omega_s n}$$

$$\omega_s = 2\pi/N$$

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$$= \sum_{k=0}^{N-1} X(k\omega) e^{+j\frac{2\pi k}{N}n} \qquad \text{Let } 2\pi/\omega_s = K$$

$$\omega_s = 2\pi/K$$

- Consider the Inverse Discrete-Time Fourier Transform....
  - What happens if we sample  $X(\omega)$ ?

$$x[n] = \sum_{k=0}^{N-1} X[k]e^{+j\frac{2\pi}{N}}$$
kn

The DTFT becomes the Discrete-Time Fourier Series