

### [11:00-11:05] Logistics for midterm #1

The exam is open book, open note, open laptop/tablet, but no networking is allowed.

If you want the ability to use MATLAB on the exam, ensure that you have a working local version installed, since you will not be able to use the web version.

The exam will cover material up to and including the 9/30 lecture. The exam will cover topics relating to the lectures, homework assignments, and mini project.

### [11:05-11:45] Sampling and aliasing

When sampling, any frequencies beyond  $(-\frac{f_s}{2}, \frac{f_s}{2})$  will alias down to a frequency within this range.

A sinusoidal signal  $x(t) = A \cos(2\pi f_0 t + \phi)$  becomes  $x[n] = \cos\left(2\pi \frac{f_0}{f_s} n\right)$  when sampled. The samples  $y[n]$  of  $y(t) = A \cos(2\pi(f_0 + \ell f_s)t + \phi)$  are identical to the samples of  $x[n]$  for any integer  $\ell$ .

The scenario where  $f_0 < f_s/2$  is called oversampling.

The scenario where  $f_0 > f_s/2$  is called undersampling.

#### Example (undersampling):

$$x(t) = \cos(2\pi f_0 t), \quad f_0 = 100 \text{ Hz}, \quad f_s = 80 \text{ Hz}$$

$$\hat{\omega}_0 = 2\pi \frac{f_0}{f_s} = 2.5\pi$$

Since  $x[n] = \cos(2.5\pi n) = \cos(0.5\pi n + 2\pi n) = \cos(0.5\pi n)$  the samples of  $\cos(2\pi 100 t)$  are the same as the samples of  $\cos(2\pi 20 t)$  when sampled at  $f_s = 80 \text{ Hz}$  (aliasing).

However, applying a standard discrete-to continuous reconstruction procedure to  $x[n]$  will result in a signal that resembles  $\cos(2\pi 20 t)$ , since  $-\frac{f_s}{2} < 20 \text{ Hz} + \ell f_s < \frac{f_s}{2}$  is only satisfied when  $\ell = 0$ .

To mitigate the effect of sampling, we can apply a low-pass analog filter (e.g. RC filter) to attenuate any frequencies above  $f_s/2$ .

#### Example (folding by undersampling)

$$x(t) = \cos(2\pi f_0 t), \quad f_0 = 100 \text{ Hz}, \quad f_s = 125 \text{ Hz}$$

$$x[n] = \cos\left(2\pi \frac{f_0}{f_s} n\right) = \cos(1.6\pi n) = \cos(1.6\pi n - 2\pi n) = \cos(-0.4\pi n)$$

### [11:40-] Reconstruction (discrete-to-continuous conversion)

The general form of interpolation is a mixed (continuous and discrete) convolution:

$$\tilde{y}(t) = \sum_{n=-\infty}^{\infty} y[n] p(t - T_s n)$$

Input: discrete-time sequence  $y[n] = y(nT_s)$

Output: continuous-time signal that is an approximation of  $y(t)$

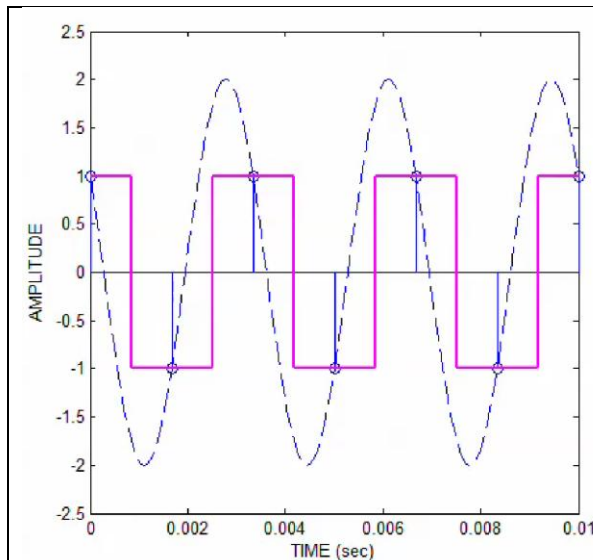
The pulse function  $p(t)$  is chosen to have unit amplitude and/or area.

**Rectangular pulse** with height 1 and width  $T_s$ .

Linear interpolation: equivalent to  $p(t) = \text{triangular pulse}$  with height 1 and width  $2T_s$ .

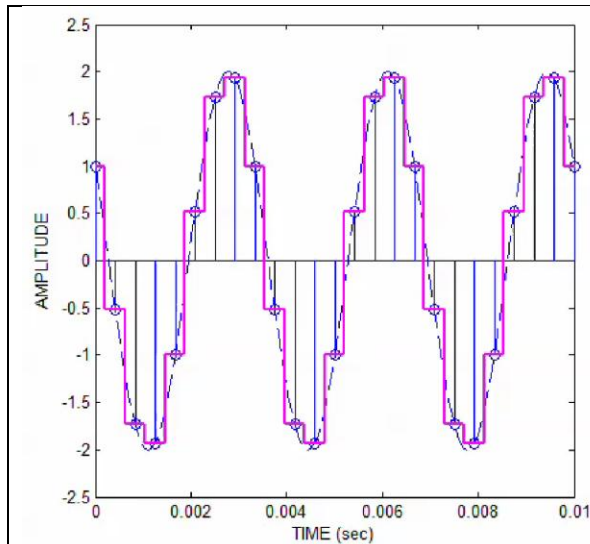
To avoid interfering with other samples,  $p(t \pm \ell T_s) = 0$  for all non-zero integers  $\ell \neq 0$ .

**Sinc pulse:**  $p(t) = \text{sinc}(t/T_s) = \frac{\sin(\pi t/T_s)}{\pi t/T_s}$ . The sinc pulse has infinite overlap with other sinc pulses, but the zero crossings occur at other sampling times to avoid interference.



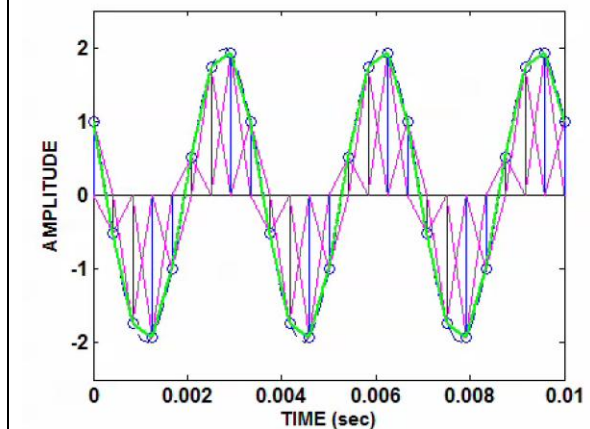
#### Reconstruction with a square pulse Sampling near the Nyquist rate

- Captures the correct number of zero crossings
- Amplitude is reduced
- Shape is not captured



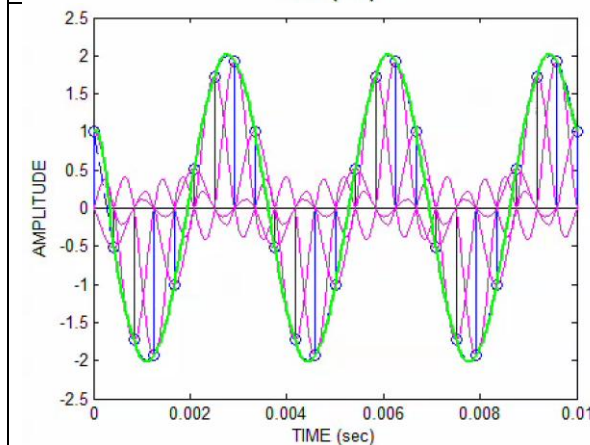
Reconstruction with a square pulse  
**Sampling at  $>8\times$  the Nyquist rate**

- Accurately tracks the amplitude
- Shape is closer to a sinusoid



Reconstruction with triangular pulse  
**Sampling at  $>8\times$  the Nyquist rate**

- Shape improves even more



Reconstruction with truncated sinc pulse  
**Sampling at  $>8\times$  the Nyquist rate**

- Shape is nearly perfect
- Amplitude is nearly perfect

### [11:10] Power consumption

In a circuit, the dynamic power can be modeled by

$$P = ACV^2 f$$

Where  $f$  is the operating frequency of the circuit.

Some data converters have power  $\propto f^2$ . Additionally, many signal processing algorithms require complexity between  $n$  and  $n^2$  in the number of samples  $n$ . Thus, oversampling can be very expensive from a power perspective.

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$x(t) \xrightarrow{T_s} x[n]$   
 $f_0 = 100 \text{ Hz}$   
 $f_s = 80 \text{ Hz}$   
 $x(t) = \cos(2\pi f_0 t)$   
 $\hat{\omega}_0 = 2\pi \frac{f_0}{f_s}$   
 $x[n] = \cos(\hat{\omega}_0 n)$

$\hat{\omega}_0 = 2\pi \frac{100 \text{ Hz}}{80 \text{ Hz}}$   
 $\hat{\omega}_0 = 2\pi(1.25)$   
 $\hat{\omega}_0 = 2.5\pi$  aliases to  $0.5\pi$

$0.5\pi = 2\pi \frac{f_{\text{aliased}}}{f_s}$   
 $2\pi(0.25) = 2\pi \frac{f_{\text{aliased}}}{80 \text{ Hz}}$   
 $f_{\text{aliased}} = (0.25)(80 \text{ Hz}) = 20 \text{ Hz}$

By sampling,  $f_s > 2f_{\text{max}} \rightarrow f_{\text{max}} < \frac{1}{2}f_s$

Sampling can only capture  $-\frac{1}{2}f_s < f < \frac{1}{2}f_s$

$\hat{\omega} = 2\pi \frac{f}{f_s}$   
 $-\pi < \hat{\omega} \leq \pi$   
 Discrete-time frequency domain is periodic with periodicity  $2\pi$

$x[n] = \cos(2.5\pi n)$   
 $x[n] = \cos(2.5\pi n - 2\pi n) = \cos(0.5\pi n)$

## Power Consumption in a Circuit

Continuous  
Time  
Circuits

$$P \propto AC V_{dd}^2 f$$

↑ activity factor

Discrete-Time Algorithms

Fast Fourier Transform (FFT)

for a block of  $N$  samples

$N \log_2 N$  for  $N$  is a power of two

$N^2$

$N$  is prime

Some A/D converters: Power  $\propto f_s^2$

Overall, the power could

increase by  $f_s$  to  $f_s^2$

for increasing  $f_s$