

Discussion 8A

Circular Convolution and Aliasing

* Circulant Matrices

- Derivation: DFT Basis Diagonalize Circulant Matrices

* Circular Convolution and DFT

- Derivation: Circular Convolution in Time is Multiplication in Frequency

* Sampling and Aliasing

(I.) Circulant Matrices

Consider a matrix as follows:

$$C_h = \begin{bmatrix} h_0 & h_{N-1} & \dots & h_2 & h_1 \\ h_1 & h_0 & h_{N-1} & \dots & h_2 \\ \vdots & & & & \\ h_{N-1} & h_{N-2} & \dots & h_1 & h_0 \end{bmatrix} \leftarrow \begin{matrix} N \times N \\ \text{matrix} \end{matrix}$$

Each row is rotated one element to the right
(as we go down the rows)

If $\vec{h} = \begin{bmatrix} h_0 \\ h_1 \\ \vdots \\ h_{N-1} \end{bmatrix}$ \leftarrow column vector, defines C_h

DFT $\vec{h} \rightarrow [H \dots]$

has $H = \begin{pmatrix} H[0] \\ \vdots \\ H[N-1] \end{pmatrix}$

Then,

$$C_u \vec{u}_k = (\sqrt{N} H[k]) \vec{u}_k$$

\uparrow eigenvec \uparrow eigenval

vector $\vec{u}_k = \frac{1}{\sqrt{N}} \begin{bmatrix} \omega_N^{k \cdot 0} \\ \omega_N^{k \cdot 1} \\ \vdots \\ \omega_N^{k \cdot N-1} \end{bmatrix}$

In other words, C_u is diagonalized by the DFT basis!

$$C_u = U \Lambda_H U^* \quad (U = [\vec{u}_0 \dots \vec{u}_{N-1}])$$

$$= F^* \Lambda_H F \quad (F = U^*)$$

$$C_u = F^* \begin{bmatrix} \sqrt{N} H[0] & & & 0 \\ & \sqrt{N} H[1] & & \\ & & \ddots & \\ 0 & & & \sqrt{N} H[N-1] \end{bmatrix} F$$

Proof?

$$h[n] = \frac{1}{\sqrt{N}} \sum_{k=0}^{N-1} H[k] \omega_N^{kn}$$

← DFT synthesis equation (summation)

$$h[n-n_0] = \frac{1}{\sqrt{N}} \sum_{k=0}^{N-1} H[k] \omega_N^{k(n-n_0)}$$

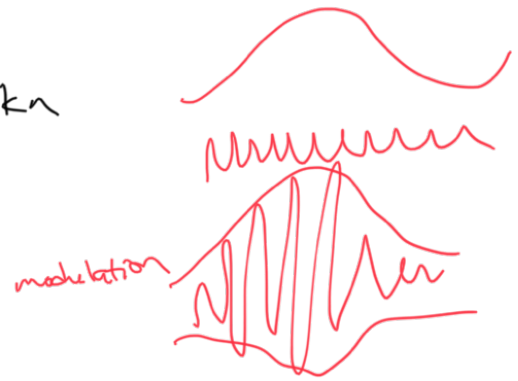
$$= \frac{1}{\sqrt{N}} \sum_{k=0}^{N-1} H[k] \omega_N^{-kn_0} \omega_N^{kn}$$

$$= \frac{1}{\sqrt{N}} \sum_{k=0}^{N-1} \hat{H}[k] \omega_N^{kn}$$



(0 shift)

$\hat{H}[k]$



modulation

↳ DFT coefficients of $h[n-n_0] = \hat{h}[k]$

are $\hat{H}[k] = H[k] \omega_N^{-kn_0}$

⇒ circular shifts in time are modulation in frequency!

$$h[n-n_0] \xleftrightarrow{\text{DFT}} H[k] \omega_N^{-kn_0}$$

Thus,

$$C_n = \begin{bmatrix} \uparrow \\ \vec{h}_0 \\ \uparrow \\ \vdots \\ \vec{h}_{N-1} \\ \uparrow \end{bmatrix} = \begin{bmatrix} h_0 & h_{N-1} & \dots & h_1 \\ h_1 & h_0 & \dots & h_2 \\ \vdots & \vdots & \dots & \vdots \\ h_{N-1} & h_{N-2} & \dots & h_0 \end{bmatrix}$$

h. indicates a circular shift

n_0 means a shift of n_0 ,
 so h_0 has 0 shift

Note that if $h[n-n_0 \pmod N] \xleftrightarrow{\text{DFT}} H[k] \omega_N^{-kn_0}$,

then $F \vec{h}_{n_0} = \sqrt{N} \vec{H} \odot \vec{u}_{n_0} = \frac{1}{\sqrt{N}} \begin{bmatrix} \omega_N^{-0 \cdot n_0} \\ \omega_N^{-1 \cdot n_0} \\ \vdots \\ \omega_N^{-(N-1) \cdot n_0} \end{bmatrix}$ element wise multiplication

$$= \sqrt{N} \frac{1}{\sqrt{N}} \begin{bmatrix} H[0] \omega_N^{-0 \cdot n_0} \\ H[1] \omega_N^{-1 \cdot n_0} \\ \vdots \\ H[N-1] \omega_N^{-(N-1) \cdot n_0} \end{bmatrix}$$

$$C_h = \begin{bmatrix} \vec{h}_0 & \vec{h}_1 & \dots & \vec{h}_{N-1} \end{bmatrix} = \begin{bmatrix} \sqrt{N} F^* (\vec{H} \odot \vec{u}_0) & \sqrt{N} F^* (\vec{H} \odot \vec{u}_1) & \dots & \sqrt{N} F^* (\vec{H} \odot \vec{u}_{N-1}) \end{bmatrix}$$

$$= F^* \begin{bmatrix} \sqrt{N} \vec{H} \odot \vec{u}_0 & \sqrt{N} \vec{H} \odot \vec{u}_1 & \dots & \sqrt{N} \vec{H} \odot \vec{u}_{N-1} \end{bmatrix}$$

$$= F^* \begin{bmatrix} \sqrt{N} H[0] & & & \\ & \sqrt{N} H[1] & & \\ & & \ddots & \\ & & & \sqrt{N} H[N-1] \end{bmatrix} \begin{bmatrix} \vec{u}_0 \\ \vec{u}_1 \\ \vdots \\ \vec{u}_{N-1} \end{bmatrix}$$

$$= F^* \begin{bmatrix} \sqrt{N} H[0] & & & 0 \\ & \sqrt{N} H[1] & & \\ & & \ddots & \\ 0 & & & \sqrt{N} H[N-1] \end{bmatrix} F \vec{u}$$

$$= F^* \Lambda_{\#} F$$

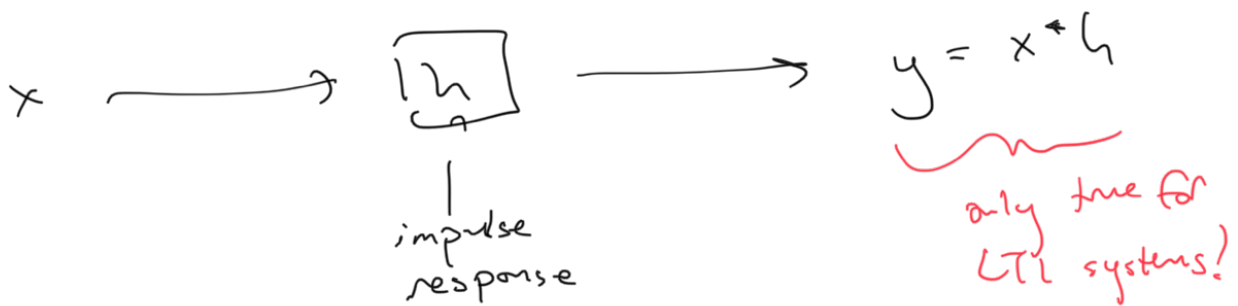
(A.) Circular Convolution and DFT

From lecture, we can describe the input output relationship of a periodic discrete-time LTI system with

$$\vec{y} = C_h \vec{x}$$

circular convolution

Remember, convolution is important because it allows us to calculate the output of a LTI system!



We will show:

"Circular convolution in time is element-wise multiplication in frequency"

$$x[n] * h[n] \xleftrightarrow{\text{DFT}} \sqrt{N} \vec{X} \odot \vec{H}$$

Proof:

$$\vec{y} = C_h \vec{x} = x[n] * h[n] \text{ (circular)}$$

y ← h

convolution

where

$$C_h \vec{s}_0 = \vec{h}$$

$$s_r[n] = \begin{cases} 1 & n=k \\ 0 & \text{else} \end{cases}$$

n=k
else

$$\vec{s}_0 = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$S = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ & 1 & & & \\ & & \dots & & \\ & & & 1 & \\ & & & & 0 & \dots & 0 \end{bmatrix}$$

$$S \begin{pmatrix} h_0 \\ \vdots \\ h_{N-1} \end{pmatrix} = \begin{pmatrix} h_1 \\ h_2 \\ \vdots \\ h_{N-1} \\ h_N \end{pmatrix}$$

circularly shifted!

Then

$$C_h = \begin{bmatrix} \vec{h}_0 & \vec{h}_1 & \dots & \vec{h}_{N-1} \\ \vdots & \vdots & & \vdots \end{bmatrix}$$

$$\vec{y} = C_h \vec{x} = F^* \Lambda_H F \vec{x}$$

$$\Rightarrow F \vec{y} = \Lambda_H F \vec{x}$$

But, $\vec{Y} = F \vec{y}$, $\vec{X} = F \vec{x}$

$$S_o \quad \vec{Y} = \Lambda_H \vec{X}$$

$$= \sqrt{N} \begin{bmatrix} H[0] & & & \\ & H[1] & & \\ & & \dots & \\ & & & H[N-1] \end{bmatrix} \begin{bmatrix} X[0] \\ \vdots \\ X[N-1] \end{bmatrix}$$

$$= \sqrt{N} \begin{bmatrix} H[0] X[0] \\ H[1] X[1] \\ \vdots \\ H[N-1] X[N-1] \end{bmatrix}$$

Or, $\vec{Y} = \sqrt{N} \vec{H} \odot \vec{X}$ ← vectors

$Y[k] = \sqrt{N} H[k] X[k]$ ← scalars
(kth element considered)

In Q1 of D.38A,
what if our system is not periodic?

⇒ zero-pad both \vec{x} and \vec{h} , where
 \vec{x} is length N and \vec{h} is length M ,
to length $M+N-1$.

Procedure

(i) Compute DFT of \vec{x} and \vec{h}
(where they are appropriately zero-padded
& not a DT periodic LTI systems)

$\vec{X} = F\vec{x}, \vec{H} = F\vec{h}$

(ii) $\vec{Y} = \sqrt{N} \vec{X} \odot \vec{H}$

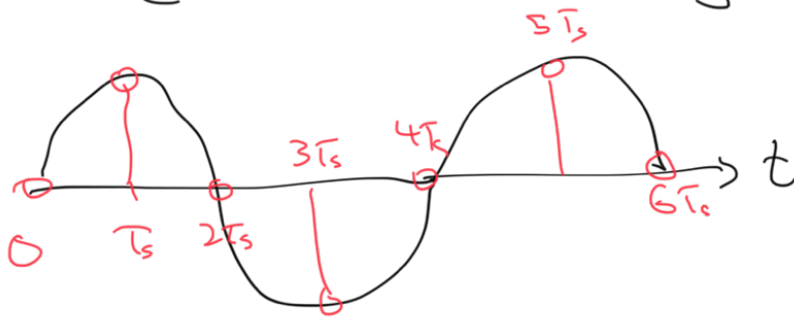
(iii) Inverse DFT to get \vec{y}

$\vec{y} = F^{-1} \vec{Y}$

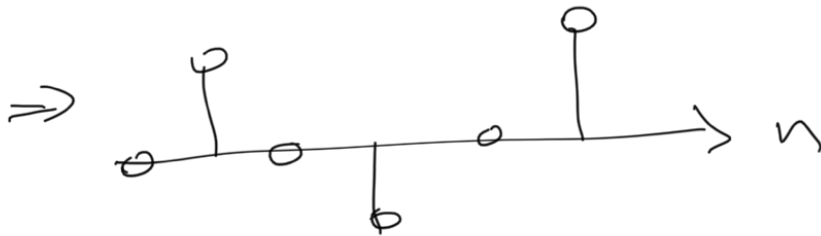
III.

Sampling Theorem

(Intro to sampling + aliasing)



Sample w/ some period T_s



How to reconstruct?

Main way: "sinc interpolation"

reconstruct via linear combination of shifted basis functions

Reconstructed signal: $\hat{x}(t) = \sum x[n] \Phi(t - nT_s)$

$$\Phi(t) = \text{sinc}\left(\frac{t}{T_s}\right)$$

Will it always perfectly reconstruct the original signal $x(t)$, i.e. will $\hat{x}(t) = x(t)$?

NO!

Depends on sampling frequency / period

— need to sample "fast enough" to capture all variations in the signal

recovered

If $T_s \equiv$ sampling period (s)

$$\omega_s = \frac{2\pi}{T_s} \equiv \text{sampling frequency } \left(\frac{\text{rad}}{\text{s}} \right)$$

Nyquist - Shannon Sampling Theorem

If the highest frequency present in the original signal is ω_{\max} , then to perfectly reconstruct we require:

$$\omega_s > 2\omega_{\max}$$

Mnemonic: human hearing $\sim 20 - 20 \text{ kHz}$

$$\Rightarrow f_{\max} = 20 \text{ kHz}$$

Music CDs (?) sample at $f_s = 44.1 \text{ kHz}$

$$\Rightarrow f_s \approx 2f_{\max}$$

Recover some other signal if $\omega_s \leq 2\omega_{\max}$



(helicopter video
— see today's lecture)

"... alias effect"

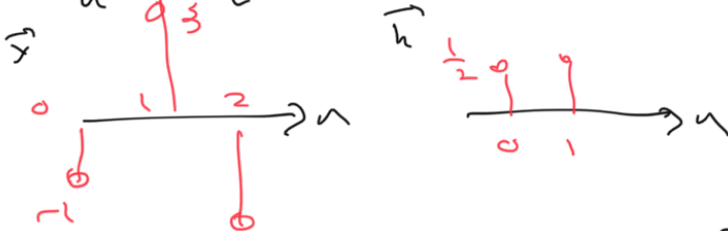
Q1, Q4

↓ ↓
 DFT aliasing
 and evolution

① Circulant Matrices and Convolution

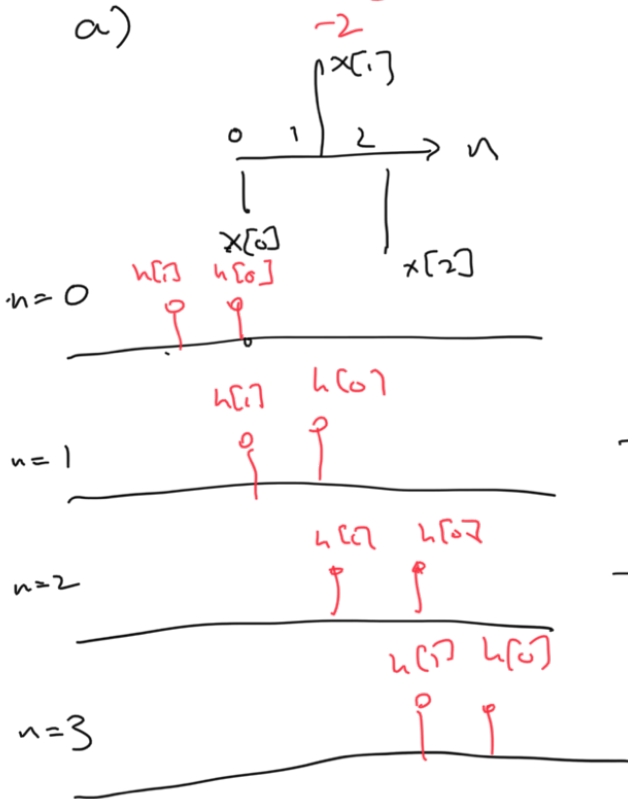
$$\vec{x} = [-1 \ 3 \ -2]^T$$

$$\vec{h} = [0.5 \ 0.5]^T$$



a)

Calculate $x[n] * h[n]$. What is the length of the output?



$$\longrightarrow y[0] = h[0]x[0] = -0.5$$

$$\longrightarrow y[1] = h[1]x[0] + h[0]x[1] = 1$$

$$\longrightarrow y[2] = h[1]x[1] + h[0]x[2] = 0.5$$

$$\longrightarrow y[3] = h[1]x[2] = -1$$

$$\vec{y} = [-0.5 \ 1 \ 0.5 \ -1]^T$$

\vec{x} : $M = 3$ (length)

\vec{h} : $N=2$ in length

\rightarrow y is $M+N-1 = 4$ in length

If \vec{x} is length M and \vec{h} is length N ,

$y = x * h$ is length $M+N-1$

\Rightarrow zero pad both x and h to be this length

b) As a matrix multiplication?

$$\begin{bmatrix} y[0] \\ y[1] \\ y[2] \\ y[3] \end{bmatrix} = \begin{bmatrix} h[0] & 0 & 0 & 0 \\ h[1] & h[0] & 0 & 0 \\ 0 & h[1] & h[0] & 0 \\ 0 & 0 & h[1] & h[0] \end{bmatrix} \begin{bmatrix} x[0] \\ x[1] \\ x[2] \\ 0 \end{bmatrix}$$

\vec{y} A \vec{x}

Almost circulant ... need to zero pad

c) Zero pad!

$$\vec{h} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & 0 & 0 \end{bmatrix}$$

$$\vec{x} = \begin{bmatrix} -1 & 3 & -2 & 0 \end{bmatrix}$$

$$\vec{y} = \begin{bmatrix} h[0] & 0 & 0 & h[1] \\ h[1] & h[0] & 0 & 0 \\ 0 & h[1] & h[0] & 0 \\ 0 & 0 & h[1] & h[0] \end{bmatrix} \begin{bmatrix} x[0] \\ x[1] \\ x[2] \\ 0 \end{bmatrix}$$

$$C_h = \begin{bmatrix} \hat{h}_0 & \hat{h}_1 & \hat{h}_2 & \hat{h}_3 \\ \vdots & \vdots & \vdots & \vdots \end{bmatrix}$$

circular shifts of \hat{h}

d) Calculate \hat{y} using the DFT instead of convolver

$$\vec{X} = F \vec{x}, \quad \vec{H} = F \vec{h}$$

$$F^* = \frac{1}{\sqrt{4}} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 0 & -1 & -j \\ 1 & -j & 1 & -1 \\ 1 & j & -1 & 1 \end{bmatrix}, \quad F = \frac{1}{\sqrt{4}} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & j \end{bmatrix}$$

How? $\vec{u}_0, \vec{u}_1, \vec{u}_2, \vec{u}_3$
for $N=4$ (length 4 signals)

$$\vec{u}_0 = \frac{1}{\sqrt{4}} \begin{bmatrix} \omega_4^{0 \cdot 0} \\ \omega_4^{0 \cdot 1} \\ \omega_4^{0 \cdot 2} \\ \omega_4^{0 \cdot 3} \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\vec{u}_1 = \frac{1}{\sqrt{4}} \begin{bmatrix} \omega_4^{1 \cdot 0} \\ \omega_4^{1 \cdot 1} \\ \omega_4^{1 \cdot 2} \\ \omega_4^{1 \cdot 3} \end{bmatrix} = \frac{1}{\sqrt{4}} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} = \frac{1}{\sqrt{4}} \begin{bmatrix} 1 \\ 0 \\ -1 \\ 1 \end{bmatrix}$$

$$\begin{aligned} \omega_4 &= e^{j \frac{2\pi}{4}} \\ &= e^{j \frac{\pi}{2}} \\ &= j \end{aligned}$$

$$\vec{u}_2 = \frac{1}{\sqrt{4}} \begin{bmatrix} (-1)^0 \\ (-1)^1 \\ (-1)^2 \\ (-1)^3 \end{bmatrix} = \frac{1}{\sqrt{4}} \begin{bmatrix} 1 \\ -1 \\ 1 \\ -1 \end{bmatrix}$$

$$\vec{u}_3 = \frac{1}{\sqrt{4}} \begin{bmatrix} (j)^0 \\ (j)^1 \\ (j)^2 \\ (j)^3 \end{bmatrix} = \frac{1}{\sqrt{4}} \begin{bmatrix} 1 \\ j \\ -1 \\ -j \end{bmatrix}$$

$$\vec{1} = \begin{bmatrix} 0 \\ \dots \\ 1 \end{bmatrix}$$

$$\hat{A} = \begin{bmatrix} 0.5 - 1.5j \\ -3 \\ 0.5 + 1.5j \end{bmatrix} \quad H = \begin{bmatrix} 0.25 - 0.25j \\ 0 \\ 0.25 + 0.25j \end{bmatrix}$$

$$\text{(ii)} \quad \vec{y} = \sqrt{4} \vec{H} \odot \vec{X}$$

$$= 2 \begin{bmatrix} 0.5 \\ 0.25 - 0.25j \\ 0 \\ 0.25 + 0.25j \end{bmatrix} \odot \begin{bmatrix} 0 \\ 0.5 - 1.5j \\ -3 \\ 0.5 + 1.5j \end{bmatrix}$$

$$= 2 \begin{bmatrix} 0.5 \cdot 0 \\ (0.25 - 0.25j)(0.5 - 1.5j) \\ 0 \cdot -3 \\ (0.25 + 0.25j)(0.5 + 1.5j) \end{bmatrix}$$

$$= \begin{bmatrix} 0 \\ -0.5 - j \\ 0 \\ -0.5 + j \end{bmatrix}$$

$$\text{(iii)} \quad \vec{y} = F^{-1} \vec{y} = \begin{bmatrix} -0.5 \\ 1 \\ 0.5 \\ -1 \end{bmatrix} \quad \text{Same as before!}$$

e) Why is this important?

FFT: $O(N \log N)$

Convolution: $O(N^2)$

Ex: 10^6 elements

$\rightarrow 10^{12}$ operations (direct convolution)

$\rightarrow 6 \times 10^6$ operations (FFT)

(4) Aliasing

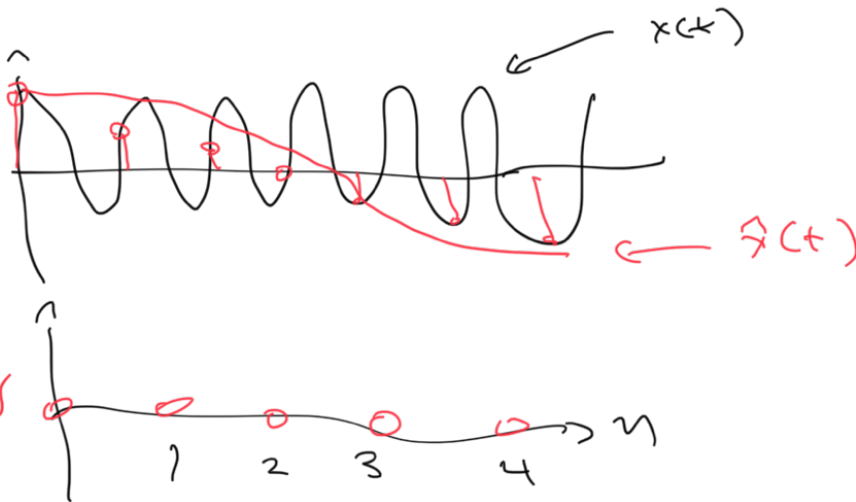
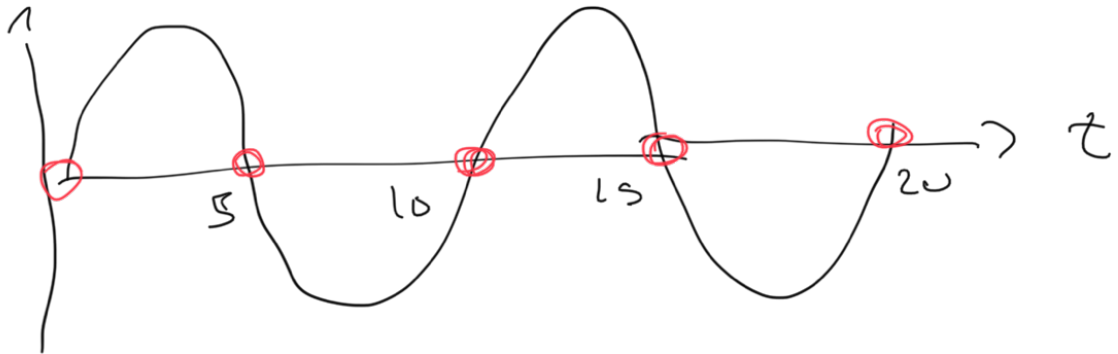
$$x(t) = \sin(0.2\pi t)$$

period

$$\sqrt{\frac{1}{T_s}}$$

be to recover

a) What would the sampling
a constant at zero?



From lecture

$$\boxed{T = 5 \text{ s}}$$

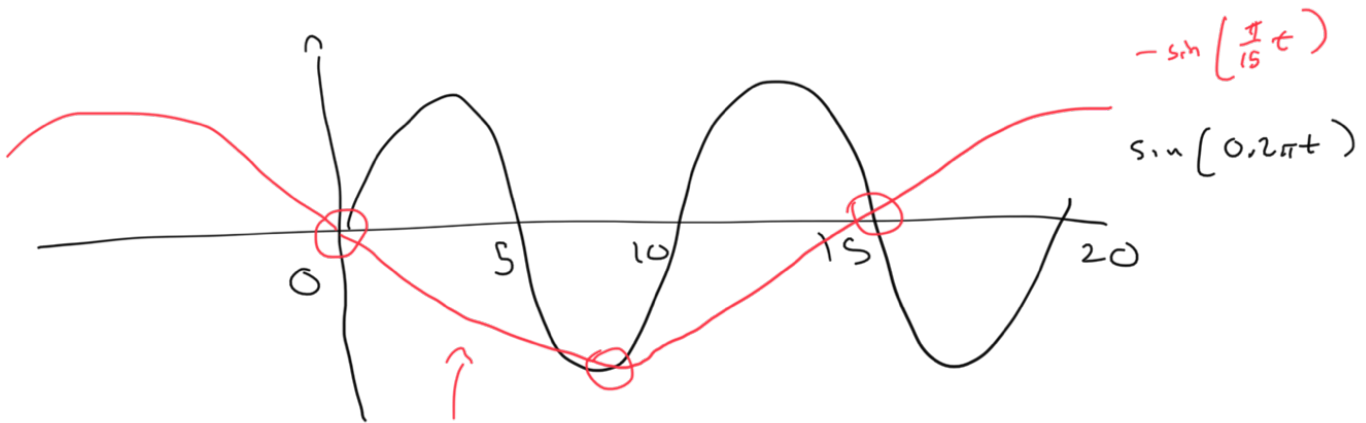
$$\sin(0.2\pi t) = 0 \quad \text{if} \quad 0.2\pi t = m\pi$$

$$t = 5m, \quad m \in \mathbb{Z}$$

$$\Rightarrow \text{sample at } T_s = 5$$

$$(\frac{\pi}{2} + \pi) ?$$

b) Recover $\hat{x}(t) = -\sin\left(\frac{\pi}{15}t\right)$



Look at intersection points — every 7.5 seconds

$$\Rightarrow \boxed{T_s = 7.5 \Delta}$$

$$x[n] = x(nT) = \hat{x}(nT)$$

$$x(t) = \sin(0.2\pi t)$$

$$\hat{x}(t) = -\sin\left(\frac{\pi}{15}t\right)$$

$$\Rightarrow \sin(0.2\pi nT) = -\sin\left(\frac{\pi}{15}nT\right)$$

$$= -\sin(-0.2\pi nT)$$

$$= -\sin(-0.2\pi nT + 2\pi n)$$

$$\begin{aligned} \sin(x) &= \sin(x + 2\pi) \\ &= \sin(x + 2\pi n) \end{aligned}$$

$$\text{Thus, } -\sin(-0.2\pi nT + 2\pi n) = -\sin\left(\frac{\pi}{15}nT\right)$$

True for all n , so

$$-0.2\pi nT + 2\pi n = \frac{\pi}{15} nT$$

$$-0.2\pi T + 2\pi = \frac{\pi}{15} T$$

$$2\pi = \frac{4\pi}{15} T$$

$$\boxed{T = 7.5}$$

Alias at $\omega \pm 2\pi i$, $i \in \mathbb{Z}$

② Sampling Theorem Basics

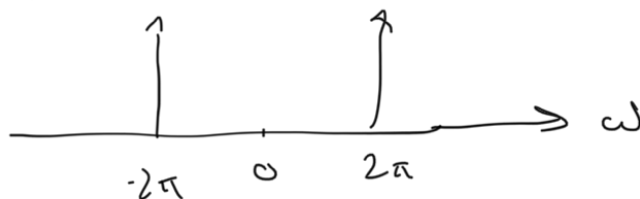
$$x(t) = \cos(2\pi t)$$

a) $\cos(\omega t) = \cos(2\pi t)$

$$\Rightarrow \begin{cases} \omega = 2\pi \text{ rad/s} = \omega_{\max} \\ f = \frac{\omega}{2\pi} = 1 \text{ Hz} = f_{\max} \end{cases}$$

not DFT!

(FT would look like this:)



b)

Sample every T seconds:

$$\boxed{\omega_s = \frac{2\pi}{T}}$$

by definition

c) Smallest T w/ imperfect reconstruction:

$\omega_s > 2\omega_{\max}$ for perfect reconstruction

$\Rightarrow \omega_s \leq 2\omega_{\max}$ for imperfect

Thus, $\frac{2\pi}{T} \leq 2\omega_{\max} \Rightarrow T \geq \frac{\pi}{\omega_{\max}}$

If $\omega_{\max} = 2\pi$,

$$T \geq \frac{\pi}{2\pi} = \frac{1}{2} \text{ s}$$

$$\boxed{T_{\min} = \frac{1}{2} \text{ s}}$$

(3) More Sampling

$$T_m = \frac{1}{4} \text{ s}, \quad T_n = 1 \text{ s}$$

\Downarrow
reconstructs f_m

\Downarrow
reconstructs f_n

original signal:

$$x(t) = \cos(2\pi t)$$

a) Nyquist limit satisfied?

$$\omega_s > 2\omega_{\max} = 2(2\pi)$$

$$\text{If } \omega_s = \frac{2\pi}{T}, \quad \frac{2\pi}{T} > 2(2\pi) \Rightarrow T < \frac{1}{2} \text{ s}$$

$T_m < \frac{1}{2} \text{ s}$	✓	satisfied
$T_n > \frac{1}{2} \text{ s}$	✗	not satisfied

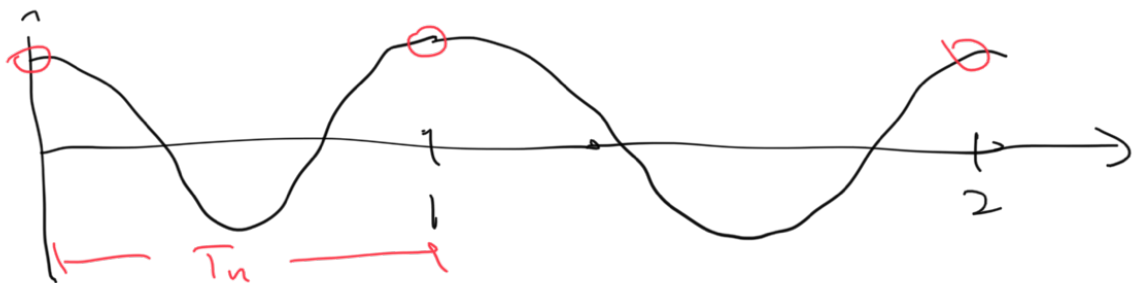
b) Reconstruct w/ T_u ?

Then $\omega_{s,n} = \frac{2\pi}{T} = 2\pi \text{ rad/s}$

$$2\omega_{\max} < \omega_{s,n}$$

$$\Rightarrow \omega_{\max} = \frac{\omega_{s,n}}{2} = \frac{2\pi}{2} = \pi \text{ rad/s}$$

c) Periodic functions w/ freq $\omega \leq \pi$,
but with same samples?



Just constant samples of 1!

$$\text{Reconstruct } f_n(t) = 1$$