#### **Circulant Matrices**

A square matrix  $C_h$  is circulant if each row vector is rotated one element to the right relative to the preceding row vector.

$$C_{h} = \begin{bmatrix} h_{0} & h_{N-1} & \cdots & h_{2} & h_{1} \\ h_{1} & h_{0} & h_{N-1} & & h_{2} \\ \vdots & h_{1} & h_{0} & \ddots & \vdots \\ h_{N-2} & \vdots & \ddots & \ddots & h_{N-1} \\ h_{N-1} & h_{N-2} & \cdots & h_{1} & h_{0} \end{bmatrix}$$
(1)

Recall from lecture that we can describe the input-output relationship of a periodic discrete-time LTI system via a circulant matrix.

$$\vec{y} = C_h \vec{x} \tag{2}$$

In this case, the first column of  $C_h$  is the impulse response h[n] of the system.

$$\vec{h} = \begin{bmatrix} h_0 & h_1 & \cdots & h_{N-2} & h_{N-1} \end{bmatrix}$$
 (3)

Rather beautifully, the DFT basis vectors are eigenvectors of  $C_h$ . We will have N DFT vectors, since that is the dimensionality of our model.

$$\vec{u_k} = \frac{1}{\sqrt{N}} \begin{bmatrix} 1 & e^{j\frac{2\pi}{N}k \cdot 1} & \cdots & e^{j\frac{2\pi}{N}k \cdot (N-1)} \end{bmatrix}$$
 (4)

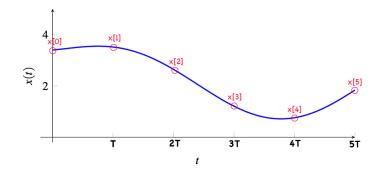
Letting H[k] be the  $k^{th}$  DFT coefficient of h[n], we can write the following eigenvalue equation for  $k = 0, 1, \dots, N - 1$ .

$$C_h \vec{u}_k = \underbrace{\left(\sqrt{N} \times H[k]\right)}_{\text{eigenvalue}} \vec{u}_k \tag{5}$$

In this discussion you'll see why this is useful by representing convolution as a circulant matrix  $C_h$ , and then diagonalizing it. This will draw the connection between the DFT and LTI systems.

## Sampling theorem

Let *x* be continuous signal bandlimited by frequency  $\omega_{max}$ . We sample *x* with a period of  $T_s$ .



Given the discrete samples, we can try reconstructing the original signal f through sincinterpolation where  $\Phi(t) = \mathrm{sinc}\left(\frac{t}{T_s}\right)$ 

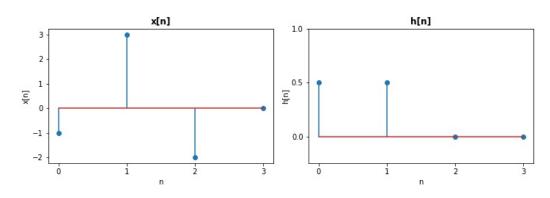
$$\hat{x}(t) = \sum_{n=-\infty}^{\infty} x[n]\Phi(t - nT_s)$$

We define the **sampling frequency** as  $\omega_s = \frac{2\pi}{T_s}$ . The Sampling Theorem says if  $\omega_{max} < \frac{\pi}{T_s}$ , or  $\omega_s > 2\omega_{max}$ , then we are able to recover the original signal, i.e.  $x = \hat{x}$ .

#### 1 Circulant Matrices & Convolution

Consider the signal x[n] of length 3 and an impulse response h[n] of length 2. You may assume that they are zero everywhere else.

$$\vec{x} = \begin{bmatrix} -1 & 3 & -2 \end{bmatrix}^T \qquad \vec{h} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \end{bmatrix}^T \tag{6}$$



- a) What is the convolution y[n] = x[n] \* h[n]? Also what is the length of this output signal?
- b) Now write each term of the output signal y[n] as a sum using the convolution formula and set up a matrix equation  $\vec{y} = A\vec{x}$ . What is the size of this matrix?
- c) Add elements to the matrix A and zeros to the vector  $\vec{x}$  to create a square matrix  $C_h$  that is circulant.
- d) Since the DFT diagonalizes circulant matrices, lets try to solve for the output signal y[n] using the DFT instead of convolution.
  - Step 1: Compute the DFT of x[n] and h[n]:  $\vec{X} = F\vec{x}$ ,  $\vec{H} = F\vec{h}$ .
  - Step 2: Take the elementwise product of the DFTs and scale:  $\vec{Y} = \sqrt{N}\vec{X}\odot\vec{H}$ .
  - Step 3: Perform the inverse DFT to get the result  $\vec{y} = F^* \vec{Y}$ .
- e) What is the importance behind this result? Compare the runtimes between convolution and the Fast Fourier Transform (FFT) which takes  $O(N \log N)$  operations.

# 2 Sampling Theorem basics

Consider the following signal, x(t) defined as,

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

- a) Find the maximum frequency,  $\omega_{\text{max}}$ , in radians per second? In Hertz? (From now on, frequencies will refer to radians per second.)
- b) If I sample every *T* seconds, what is the sampling frequency?
- c) What is the smallest sampling period *T* that would result in an imperfect reconstruction?

# 3 More Sampling

Let's sample the signal from the previous question x with sampling period  $T_m = \frac{1}{4}s$  and  $T_n = 1s$  and perform sinc interpolation on the resulting samples. Let the reconstructed functions be  $f_m$  and  $f_n$ .

- a) Have we satisfied the Nyquist limit (i.e. the sampling theorem) in any case?
- b) What is the highest frequency we can reconstruct with the sampling rate  $T_n$ ?
- c) Based on this answer, can you think of any periodic function that has a frequencies less than or equal to  $\pi$  that samples the same as  $f_n$ ?

### 4 Aliasing

Consider the signal  $x(t) = \sin(0.2\pi t)$ .

- a) At what period T should we sample so that sinc interpolation recovers a function that is identically zero?
- b) At what period T can we sample at so that sinc interpolation recovers the function  $f(t) = -\sin\left(\frac{\pi}{15}t\right)$ ?