

## 1 Circulant Matrices

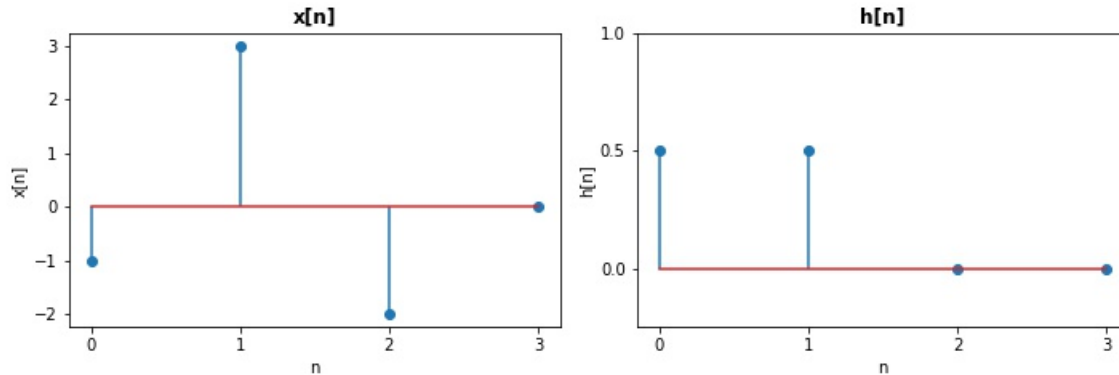
A square matrix  $C_h$  is circulant if each row vector is rotated one element to the right relative to the preceding row vector.

$$C_h = \begin{bmatrix} h_0 & h_{N-1} & \cdots & h_2 & h_1 \\ h_1 & h_0 & h_{N-1} & & h_2 \\ \vdots & h_1 & h_0 & \ddots & \vdots \\ h_{N-2} & \vdots & \ddots & \ddots & h_{N-1} \\ h_{N-1} & h_{N-2} & \cdots & h_1 & h_0 \end{bmatrix} \quad (1)$$

## 2 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.

$$\vec{x} = [-1 \quad 3 \quad -2]^T \quad \vec{h} = \left[\frac{1}{2} \quad \frac{1}{2}\right]^T \quad (2)$$



- What is the convolution  $y[n] = x[n] * h[n]$ ? Also what is the length of this output signal?
- 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?
- Add elements to the matrix  $A$  and zeros to the vector  $\vec{x}$  to create a square matrix  $C_h$  that is circulant.
- 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.

### 3 Complex Inner Product

For the complex vector space  $\mathbb{C}^n$ , we can no longer use our conventional real dot product as a valid inner product for  $\mathbb{C}^n$ . This is because the real dot product is no longer positive-definite for complex vectors. For example, let  $\vec{v} = \begin{bmatrix} j \\ j \end{bmatrix}$ . Then,  $\vec{v} \cdot \vec{v} = j^2 + j^2 = -2 < 0$ .

Therefore, for two vectors  $\vec{u}, \vec{v} \in \mathbb{C}^n$ , we define the complex inner product to be:

$$\langle \vec{u}, \vec{v} \rangle = \sum_{i=1}^n \bar{u}_i v_i = \bar{\vec{u}}^T \vec{v} = \vec{u}^* \vec{v}$$

where  $\bar{\cdot}$  denotes the complex conjugate and  $\vec{\cdot}^*$  denotes the complex conjugate transpose.

The complex inner product satisfies the following properties:

- Conjugate Symmetry:  $\langle \vec{u}, \vec{v} \rangle = \overline{\langle \vec{v}, \vec{u} \rangle}$
- Scaling:  $\langle \vec{u}, c\vec{v} \rangle = c\langle \vec{u}, \vec{v} \rangle$  and  $\langle c\vec{u}, \vec{v} \rangle = \bar{c}\langle \vec{u}, \vec{v} \rangle$
- Additivity:  $\langle \vec{u} + \vec{v}, \vec{w} \rangle = \langle \vec{u}, \vec{w} \rangle + \langle \vec{v}, \vec{w} \rangle$  and  $\langle \vec{u}, \vec{v} + \vec{w} \rangle = \langle \vec{u}, \vec{v} \rangle + \langle \vec{u}, \vec{w} \rangle$
- Positive-definite:  $\langle \vec{u}, \vec{u} \rangle \geq 0$  with  $\langle \vec{u}, \vec{u} \rangle = 0$  if and only if  $\vec{u} = \vec{0}$

Recall that the complex conjugate of a complex number  $z = a + jb = re^{j\theta}$  is equal to  $\bar{z} = a - jb = re^{-j\theta}$ .

The conjugate transpose of a vector  $\vec{v} = \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix}$  is  $\vec{v}^* = [\bar{v}_1 \quad \cdots \quad \bar{v}_n]$ .

#### Adjoint of a Matrix

The **adjoint** or **conjugate-transpose** of a matrix  $A$  is the matrix  $A^*$  such that  $A_{ij}^* = \overline{A_{ji}}$ . From the complex inner product, one can show that

$$\langle A\vec{x}, \vec{y} \rangle = \langle \vec{x}, A^*\vec{y} \rangle \quad (3)$$

A matrix is **self-adjoint** or **Hermitian** if  $A = A^*$ . Self-adjoint matrices are the complex extension of real-symmetric matrices. There is an equivalent version of the **Spectral Theorem** for such self-adjoint matrices.

#### Unitary Matrices

A **unitary** matrix is a square matrix whose columns are orthonormal with respect to the complex inner product.

$$U = [\vec{u}_1 \quad \vec{u}_2 \quad \cdots \quad \vec{u}_n], \quad \vec{u}_i^* \vec{u}_j = \begin{cases} 1, & \text{if } i = j \\ 0, & \text{otherwise} \end{cases}$$

Note that  $U^*U = UU^* = I$ , so the inverse of a unitary matrix is its conjugate transpose  $U^{-1} = U^*$ .

Since the columns of a unitary matrix are orthonormal vectors, we can interpret these matrices as “rotation” and “reflection” matrices of the standard axes. This also implies that  $\|U\vec{v}\| = \|\vec{v}\|$  for any vector  $\vec{v}$ .

## 4 Everything is Complex

In this question, we explore the similarities and differences between real and complex vector spaces.

- a) Suppose that we are given a matrix  $U$  that has orthonormal columns.

$$U = \begin{bmatrix} | & & | \\ \vec{u}_1 & \cdots & \vec{u}_n \\ | & & | \end{bmatrix} \quad (4)$$

We would like to change coordinates to express  $\vec{x}$  as a linear combination of these orthonormal basis vectors. Find scalars  $\alpha_i$  in terms of  $\vec{x}$  and  $\vec{u}_i$  so that

$$\vec{x} = \alpha_1 \vec{u}_1 + \dots + \alpha_n \vec{u}_n \quad (5)$$

- b) Show that the eigenvalues  $\lambda$  of a unitary matrix  $U$  must always have magnitude 1.
- c) Let  $A$  be an  $m \times n$  matrix with complex entries. Show that  $A^*A$  must have non-negative eigenvalues. *Hint: Think back to what we did for the real case. Consider the quantity  $\|A\vec{v}\|^2$  where  $\vec{v}$  is an eigenvector of  $A^*A$ .*