

Signal Processing and Linear Systems1

Lecture 9: Properties of the DFT

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Outline

Background

Theorems of the DFT

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L2 Norm

$$\|a\|_2 = \sqrt{a_1^2 + a_2^2 + \cdots + a_n^2}$$

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Kronecker Delta Vector

$$\delta = (1, 0, 0, \dots, 0)$$

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Matrix Transpose

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1N} \\ a_{21} & a_{22} & & a_{2N} \\ \vdots & & \ddots & \vdots \\ a_{M1} & a_{M2} & & a_{MN} \end{bmatrix}^T = \begin{bmatrix} a_{11} & a_{21} & \cdots & a_{M1} \\ a_{12} & a_{22} & & a_{M2} \\ \vdots & & \ddots & \vdots \\ a_{1N} & a_{2N} & & a_{MN} \end{bmatrix}$$

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Matrix Conjugate-Transpose

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1N} \\ a_{21} & a_{22} & & a_{2N} \\ \vdots & & \ddots & \vdots \\ a_{M1} & a_{M2} & & a_{MN} \end{bmatrix}^* = \begin{bmatrix} \bar{a}_{11} & \bar{a}_{21} & \cdots & \bar{a}_{M1} \\ \bar{a}_{12} & \bar{a}_{22} & & \bar{a}_{M2} \\ \vdots & & \ddots & \vdots \\ \bar{a}_{1N} & \bar{a}_{2N} & & \bar{a}_{MN} \end{bmatrix}$$

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Circular Convolution

Suppose $f, g \in \mathbb{C}^N$. Then the circular convolution of f and g is

$$f \circledast g = \sum_{i=1}^N f_i \tau_i \{g\}$$

where τ_i is the circular shift operator of i elements.

$$(f \circledast g)[n] = \sum_{m=1}^N f[m] g[(n - m)_{\text{mod } N}]$$

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Recall: for a discrete LSI system S

$$S\{x\} = \sum_{i=1}^N x_i \tau_i \{h\} = x \circledast h.$$

That is, the output equals the input circularly convolved with the impulse response.

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Discrete Fourier Transform

Analysis Equations:

$$\text{DFT}\{\mathbf{f}\}[m] = \sum_{n=0}^{N-1} \mathbf{f}[n] \exp\left(-i 2\pi \frac{mn}{N}\right)$$

Synthesis Equations:

$$\text{IDFT}\{\mathbf{F}\}[n] = \frac{1}{N} \sum_{m=0}^{N-1} \mathbf{F}[m] \exp\left(i 2\pi \frac{mn}{N}\right)$$

Note: both the DFT and the IDFT can be done by a computer.

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Outline

Background

Theorems of the DFT

$$\text{DFT}\{\delta\} = 1$$

Proof:

$$\begin{aligned}\text{DFT}\{\delta\}[m] &= \sum_{n=0}^{N-1} \delta \exp\left(-i 2\pi \frac{mn}{N}\right) \\ &= \exp\left(-i 2\pi \frac{0m}{N}\right) = 1.\end{aligned}$$

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The DFT Matrix

$$\text{DFT} : \mathbb{C}^N \rightarrow \mathbb{C}^N$$

The Discrete Fourier Transform is a linear transformation from a finite dimensional vector space to a finite dimensional vector space.

Therefore, it can be represented by a matrix: W . $\text{DFT}\{\mathbf{f}\} = W \mathbf{f}$

Similarly, the Inverse Discrete Fourier Transform can be represented by: W^{-1}

$$W^{-1} = \frac{1}{N} W^* \quad \text{IDFT}\{\mathbf{F}\} = W^{-1} \mathbf{F} = \frac{1}{N} W^* \mathbf{F}$$

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The Unitary DFT Matrix

$$W^{-1} = \frac{1}{N} W^*$$

$$\text{Let } W_U = \frac{1}{\sqrt{N}} W.$$

$$\text{Then } W_U^{-1} = W_U^*.$$

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Parseval's Theorem

Claim: if U is a unitary matrix, then $\|Ux\|_2 = \|x\|_2$.

Proof:

$$\|Ux\|_2 = \|x\|_2 \Leftrightarrow \|Ux\|_2^2 = \|x\|_2^2.$$

$$\|Ux\|_2^2 = (Ux)^*(Ux) = x^* U^* U x = x^* I x = x^* x = \|x\|_2^2.$$

Corollary: $\|\text{DFT}\{x\}\|_2 = \sqrt{N} \|x\|_2$.

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Convolution Theorem

$$\text{DFT}\{f \circledast g\} = \text{DFT}\{f\} \odot \text{DFT}\{g\}$$

$$\text{DFT}\{f \odot g\} = \text{DFT}\{f\} \circledast \text{DFT}\{g\}$$

Proof: see homework.

For an LSI system, $y = S\{x\}$,

$$Y[m] = X[m] H[m]$$

The DFT of the output equals the DFT of the input times the transfer function (the DFT of the impulse response).

Shift Theorem

$$\text{DFT}\{\tau_{\Delta} \mathbf{f}\}[m] = \exp\left(-i 2\pi \frac{m \Delta}{N}\right) \text{DFT}\{\mathbf{f}\}[m]$$

Proof:

$$\begin{aligned} \text{DFT}\{\tau_{\Delta} \mathbf{f}\}[m] &= \sum_{n=0}^{N-1} \mathbf{f}[n - \Delta] \exp\left(-i 2\pi \frac{mn}{N}\right) \\ &= \exp\left(-i 2\pi \frac{m \Delta}{N}\right) \sum_{\tilde{n}=0}^{N-1} \mathbf{f}[\tilde{n}] \exp\left(-i 2\pi \frac{m \tilde{n}}{N}\right) \\ &= \exp\left(-i 2\pi \frac{m \Delta}{N}\right) \text{DFT}\{\mathbf{f}\}[m]. \end{aligned}$$

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Periodicity of the DFT

So far, we've been considering the DFT as a function that maps \mathbb{C}^N to \mathbb{C}^N .

But there's nothing stopping us from computing the value of the DFT for frequencies outside of m in $\{0, \dots, N-1\}$.

When we do that (when we think of the domain of the DFT as all integers), we see that the DFT is periodic with fundamental period N .

So we can use any N consecutive values to reconstruct the input vector.

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Symmetry of the DFT

DFT{ real and even } = real and even

DFT{ real and odd } = imaginary and odd

DFT{ real } = Hermitian

Note: even and odd are defined according to the periodic extensions of the vectors.