

# The Discrete Fourier Transform

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# Discrete-Time Signals and Systems

## Reference:

Sections 7.1-7.2 of

John G. Proakis and Dimitris G. Manolakis, *Digital Signal Processing: Principles, Algorithms, and Applications*, 4th edition, 2007.

# Discrete Fourier Transform

- ▶ Frequency analysis of discrete-time signals is conveniently performed on a computer.

- ▶ Recall:

$$\begin{array}{ccc} \text{aperiodic in time} & \xleftrightarrow{\mathcal{F}} & \text{continuous in frequency} \\ x(n) & \xleftrightarrow{\mathcal{F}} & X(\omega) \end{array}$$

- ▶  $X(\omega)$  must, therefore, be stored in samples on a computer.

- ▶ What happens when we sample in the frequency domain?

# Frequency Domain Sampling

- ▶ Recall,

$$\begin{aligned} x(n) &= \frac{1}{2\pi} \int_{-\pi}^{\pi} X(\omega) e^{j\omega n} d\omega \\ X(\omega) &= \sum_{n=-\infty}^{\infty} x(n) e^{-j\omega n} \end{aligned}$$

- ▶ Suppose we sample  $X(\omega)$ .
  - ▶ Since  $X(\omega)$  is periodic with period  $2\pi$ , only a finite (say,  $N$ ) consecutive samples are needed.
  - ▶ For convenience, we consider the  $N$  equidistant samples in the interval  $0 \leq \omega \leq 2\pi$  with spacing  $\delta\omega = \frac{2\pi}{N}$ .

## Frequency Domain Sampling

$$\begin{aligned}
 X(\omega) &= \sum_{n=-\infty}^{\infty} x(n)e^{-j\omega n} \\
 X\left(\frac{2\pi}{N}k\right) &= \sum_{n=-\infty}^{\infty} x(n)e^{-j\frac{2\pi}{N}kn}, \\
 &= \sum_{l=-\infty}^{\infty} \sum_{n=IN}^{IN+N-1} x(n)e^{-j2\pi k\frac{n}{N}} \quad \text{Let } n' = n - IN \\
 &= \sum_{l=-\infty}^{\infty} \sum_{n'=0}^{N-1} x(n'+IN) \underbrace{e^{-j2\pi k\frac{n'+IN}{N}}}_{=e^{-j2\pi k\frac{n'}{N}}e^{-j2\pi k\frac{IN}{N}}} \\
 & \qquad \qquad \qquad =1
 \end{aligned}$$

## Frequency Domain Sampling

For  $k = 0, 1, 2, \dots, N-1$ ,

$$\begin{aligned}
 X\left(\frac{2\pi}{N}k\right) &= \sum_{l=-\infty}^{\infty} \sum_{n=0}^{N-1} x(n+IN)e^{-j2\pi k\frac{n}{N}} \\
 &= \sum_{n=0}^{N-1} \sum_{l=-\infty}^{\infty} x(n+IN)e^{-j2\pi k\frac{n}{N}} \\
 &= \sum_{n=0}^{N-1} \underbrace{\left[ \sum_{l=-\infty}^{\infty} x(n+IN) \right]}_{\text{equivalent signal } x_p(n)} e^{-j2\pi k\frac{n}{N}}
 \end{aligned}$$

## Frequency Domain Sampling

For  $k = 0, 1, 2, \dots, N-1$ ,

$$\begin{aligned}
 X\left(\frac{2\pi}{N}k\right) &= \sum_{n=0}^{N-1} \left[ \sum_{l=-\infty}^{\infty} x(n+IN) \right] e^{-j2\pi k\frac{n}{N}} \\
 &= \sum_{n=0}^{N-1} x_p(n) e^{-j2\pi k\frac{n}{N}}
 \end{aligned}$$

- ▶ Looks like a DTFS of  $x_p(n)$ !
- ▶ Characteristics of  $x_p(n)$ :
  - ▶ periodic
  - ▶ period =  $N$
  - ▶ can be expanded via a DTFS

## Frequency Domain Sampling

DTFS Pair:

$$\begin{aligned}
 x_p(n) &= \sum_{k=0}^{N-1} c_k e^{j2\pi k\frac{n}{N}} \\
 c_k &= \frac{1}{N} \sum_{n=0}^{N-1} x_p(n) e^{-j2\pi k\frac{n}{N}}
 \end{aligned}$$

Comparing to:

$$X\left(\frac{2\pi}{N}k\right) = \sum_{n=0}^{N-1} x_p(n) e^{-j2\pi k\frac{n}{N}}$$

## Frequency Domain Sampling and Reconstruction

Therefore,

$$c_k = \frac{1}{N} X\left(\frac{2\pi}{N}k\right) \quad k = 0, 1, \dots, N-1$$

Since,

$$x_p(n) = \sum_{k=0}^{N-1} c_k e^{j2\pi k \frac{n}{N}} \quad \text{then}$$

$$x_p(n) = \frac{1}{N} \sum_{k=0}^{N-1} X\left(\frac{2\pi}{N}k\right) e^{j2\pi k \frac{n}{N}} \quad n = 0, 1, \dots, N-1$$

## Frequency Domain Sampling and Reconstruction

Therefore,

$$x_p(n) = \frac{1}{N} \sum_{k=0}^{N-1} X\left(\frac{2\pi}{N}k\right) e^{j2\pi k \frac{n}{N}} \quad n = 0, 1, \dots, N-1$$

- **Implication:** The **samples** of  $X(\omega)$  can be used to reconstruct  $x_p(n)$ .

and since,

$$X\left(\frac{2\pi}{N}k\right) = \sum_{n=0}^{N-1} x_p(n) e^{-j2\pi k \frac{n}{N}} \quad k = 0, 1, \dots, N-1$$

- **Implication:** The signal  $x_p(n)$  can be used to reconstruct **samples** of  $X(\omega)$ .

## Frequency Domain Sampling and Reconstruction

$$\begin{aligned} x(n) &\xleftrightarrow{\mathcal{F}} X(\omega) \\ x_p(n) &\xleftrightarrow{\mathcal{F}} X\left(\frac{2\pi}{N}k\right) \end{aligned}$$

- **FACT:** We can reconstruct  $x(n)$  from  $X(\omega)$ .
- **FACT:** We can reconstruct  $x_p(n)$  from **samples** of  $X(\omega)$ .  
(... and vice versa)

- **Q:** Can we reconstruct  $x(n)$  from the **samples** of  $X(\omega)$ ?  
►  $x(n)$  Can we reconstruct  $x(n)$  from  $x_p(n)$ ?

- **A:** Maybe.

See [Figure 7.1.2 of text](#).

## Frequency Domain Sampling and Reconstruction

- $x(n)$  can be recovered from  $x_p(n)$  if there is no overlap when taking the periodic extension.
- If  $x(n)$  is finite duration and non-zero in the interval  $0 \leq n \leq L-1$ , then

$$x(n) = x_p(n), \quad 0 \leq n \leq N-1 \quad \text{when } N \geq L$$

- If  $N < L$  then,  $x(n)$  cannot be recovered from  $x_p(n)$ .  
► or equivalently  $X(\omega)$  cannot be recovered from its samples  $X\left(\frac{2\pi}{N}k\right)$  due to time-domain aliasing

Reconstruction,  $N \geq L$ 

► One way to reconstruct  $X(\omega)$  from its samples  $X\left(\frac{2\pi}{N}k\right)$ :

1. Compute  $x_p(n)$  from  $X\left(\frac{2\pi}{N}k\right)$ :

$$x_p(n) = \frac{1}{N} \sum_{k=0}^{N-1} X\left(\frac{2\pi}{N}k\right) e^{-j2\pi k \frac{n}{N}} \quad n = 0, 1, \dots, N-1$$

2. Compute  $x(n)$  from  $x_p(n)$ :

$$x(n) = \begin{cases} x_p(n) & 0 \leq n \leq N-1 \\ 0 & \text{elsewhere} \end{cases}$$

3. Compute  $X(\omega)$  from  $x(n)$ :

$$X(\omega) = \sum_{n=-\infty}^{\infty} x(n) e^{-j\omega n}$$

Reconstruction,  $N \geq L$ 

► Another way to reconstruct  $X(\omega)$  from its samples  $X\left(\frac{2\pi}{N}k\right)$ :

$$\begin{aligned} X(\omega) &= \sum_{n=-\infty}^{\infty} x(n) e^{-j\omega n} \\ &= \sum_{n=0}^{N-1} x_p(n) e^{-j\omega n} \\ &= \sum_{n=0}^{N-1} \frac{1}{N} \sum_{k=0}^{N-1} X\left(\frac{2\pi}{N}k\right) e^{j2\pi k \frac{n}{N}} e^{-j\omega n} \\ &= \sum_{k=0}^{N-1} X\left(\frac{2\pi}{N}k\right) \left[ \frac{1}{N} \sum_{n=0}^{N-1} e^{-j(\omega - 2\pi k) \frac{n}{N}} \right] \end{aligned}$$

Reconstruction,  $N \geq L$ 

$$X(\omega) = \sum_{k=0}^{N-1} X\left(\frac{2\pi}{N}k\right) \underbrace{\left[ \frac{1}{N} \sum_{n=0}^{N-1} e^{-j(\omega - 2\pi k) \frac{n}{N}} \right]}_{\text{interpolation function}}$$

$$\text{Let } P(\omega) = \frac{1}{N} \sum_{n=0}^{N-1} e^{-j\omega n} = \frac{\sin\left(\frac{\omega N}{2}\right)}{N \sin\left(\frac{\omega}{2}\right)} e^{-j\omega \left(\frac{N-1}{2}\right)}$$

$$\text{Then } X(\omega) = \sum_{k=0}^{N-1} X\left(\frac{2\pi}{N}k\right) P\left(\omega - \frac{2\pi}{N}k\right) \quad N \geq L$$

See [Figure 7.1.3 of text](#).

## Reconstruction

See [Figure 7.1.4 of text](#).

## The Discrete Fourier Transform

### Summary:

- ▶ If  $x(n)$  is infinite duration or has length  $L > N$ , the samples  $X\left(\frac{2\pi k}{N}\right)$ ,  $k = 0, 1, \dots, N - 1$  **do not** uniquely represent the original sequence  $x(n)$ .
  - ▶ Instead the frequency samples correspond to a periodic sequence  $x_p(n)$  of period  $N$  where  $x_p(n)$  is a time-aliased version of  $x(n)$ :

$$x_p(n) = \sum_{l=-\infty}^{\infty} x(n - lN)$$

## The Discrete Fourier Transform

### Summary:

- ▶ If  $x(n)$  is infinite duration or has length  $L \leq N$ , the samples  $X\left(\frac{2\pi k}{N}\right)$ ,  $k = 0, 1, \dots, N - 1$  uniquely represent the original sequence  $x(n)$ .
  - ▶ When  $x(n)$  is finite duration of length  $L \leq N$ , then  $x_p(n)$  is a periodic repetition of  $x(n)$  that can be recovered from a single period of  $x_p(n)$  using:

$$x_p(n) = \begin{cases} x(n) & 0 \leq n \leq L - 1 \\ 0 & L \leq n \leq N - 1 \end{cases}$$

- ▶ Let  $X(k) \equiv X\left(\frac{2\pi k}{N}\right)$ .

## The Discrete Fourier Transform Pair

- ▶ DFT and inverse-DFT (IDFT):

$$X(k) = \sum_{n=0}^{N-1} x(n) e^{-j2\pi k \frac{n}{N}}, \quad k = 0, 1, \dots, N - 1$$

$$x(n) = \frac{1}{N} \sum_{k=0}^{N-1} X(k) e^{j2\pi k \frac{n}{N}}, \quad n = 0, 1, \dots, N - 1$$

## DFT Example

**Q:** Determine the  $N$ -point DFT of the following sequence for  $N \geq L$ :

$$x(n) = \begin{cases} 1 & 0 \leq n \leq L - 1 \\ 0 & \text{otherwise} \end{cases}$$

**A:** The DTFT of  $x(n)$  is given by:

$$\begin{aligned} X(\omega) &= \sum_{n=0}^{L-1} x(n) e^{-j\omega n} \\ &= \sum_{n=0}^{L-1} e^{-j\omega n} = \frac{\sin(\omega L/2)}{\sin(\omega/2)} e^{-j\omega(L-1)/2} \end{aligned}$$

See [▶ Figure 7.1.5 of text](#).

## DFT Example

Thus,

$$\begin{aligned} X(\omega) &= \frac{\sin(\omega L/2)}{\sin(\omega/2)} e^{-j\omega(L-1)/2} \\ X(k) &= \frac{\sin(\frac{2\pi k}{N} L/2)}{\sin(\frac{2\pi k}{N}/2)} e^{-j\frac{2\pi k}{N}(L-1)/2} \\ &= \frac{\sin(\pi k L/N)}{\sin(\pi k/N)} e^{-j\pi k(L-1)/N} \end{aligned}$$

See [Figure 7.1.6 of text](#). See [Figure 7.1.6b of text](#).

## The DFT as a Linear Transform

$$\begin{aligned} X(k) &= \sum_{n=0}^{N-1} x(n) e^{-j2\pi k \frac{n}{N}}, \quad k = 0, 1, \dots, N-1 \\ x(n) &= \frac{1}{N} \sum_{k=0}^{N-1} X(k) e^{j2\pi k \frac{n}{N}}, \quad n = 0, 1, \dots, N-1 \end{aligned}$$

Want to convert to **matrix-vector** representation.

## The DFT as a Linear Transform

$$\begin{aligned} X(k) &= \sum_{n=0}^{N-1} x(n) e^{-j2\pi k \frac{n}{N}}, \quad k = 0, 1, \dots, N-1 \\ x(n) &= \frac{1}{N} \sum_{k=0}^{N-1} X(k) e^{j2\pi k \frac{n}{N}}, \quad n = 0, 1, \dots, N-1 \end{aligned}$$

Let  $W_N = e^{-j2\pi/N}$

$$\begin{aligned} X(k) &= \sum_{n=0}^{N-1} x(n) W_N^{kn}, \quad k = 0, 1, \dots, N-1 \\ x(n) &= \frac{1}{N} \sum_{k=0}^{N-1} X(k) W_N^{-kn}, \quad n = 0, 1, \dots, N-1 \end{aligned}$$

## The DFT as a Linear Transform

$$\begin{aligned} X(k) &= \sum_{n=0}^{N-1} x(n) W_N^{kn}, \quad k = 0, 1, \dots, N-1 \\ x(n) &= \frac{1}{N} \sum_{k=0}^{N-1} X(k) W_N^{-kn}, \quad n = 0, 1, \dots, N-1 \end{aligned}$$

$$\begin{aligned} \mathbf{x}_N &= [x(0) \ x(1) \ \dots \ x(N-1)]^T \\ \mathbf{X}_N &= [X(0) \ X(1) \ \dots \ X(N-1)]^T \\ \mathbf{W}_N &= \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & W_N & W_N^2 & \dots & W_N^{N-1} \\ 1 & W_N^2 & W_N^4 & \dots & W_N^{2(N-1)} \\ \vdots & \vdots & \vdots & \dots & \vdots \\ 1 & W_N^{N-1} & W_N^{2(N-1)} & \dots & W_N^{(N-1)(N-1)} \end{bmatrix} \end{aligned}$$

## The DFT as a Linear Transform

- In matrix-vector notation:

$$\begin{aligned}\mathbf{X}_N &= \mathbf{W}_N \mathbf{x}_N \\ \mathbf{x}_N &= \mathbf{W}_N^{-1} \mathbf{X}_N = \frac{1}{N} \mathbf{W}_N^* \mathbf{X}_N\end{aligned}$$

where it can be shown that

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

- Complexity:**  $N$  complex multiplications and  $N - 1$  complex additions;
  - For an  $N$ -point DFT, a total of  $N^2$  complex multiplications and  $N(N - 1)$  complex additions are required.

## Periodicity and Linearity

Notation:  $x(n) \xleftrightarrow{N\text{-DFT}} X(k)$

- Periodicity:

$$\begin{aligned}x(n + N) &= x(n) \quad \text{for all } n \\ X(k + N) &= X(k) \quad \text{for all } k\end{aligned}$$

- Linearity: If

$$\begin{aligned}x_1(n) &\xleftrightarrow{N\text{-DFT}} X_1(k) \\ x_2(n) &\xleftrightarrow{N\text{-DFT}} X_2(k)\end{aligned}$$

Then  $a_1 x_1(n) + a_2 x_2(n) \xleftrightarrow{N\text{-DFT}} a_1 X_1(k) + a_2 X_2(k)$

## Circular Symmetry and Convolution

- Circular operations:** apply the transformation on the **periodic repetition** of  $x(n)$  and then obtain the final result by taking points for  $n = 0, 1, \dots, N - 1$
- Circular Symmetry:**
  - circular time reversal:  $x((-n))_N = x(N - n)$
  - circularly even:  $x(N - n) = x(n)$
  - circularly odd:  $x(N - n) = -x(n)$
- Circular Convolution:**

$$x_3(m) = \sum_{n=0}^{N-1} x_1(n) x_2((m - n))_N, \quad m = 0, 1, \dots, N - 1$$

## Important DFT Properties

Property	Time Domain	Frequency Domain
Notation:	$x(n)$	$X(k)$
Periodicity:	$x(n) = x(n + N)$	$X(k) = X(k + N)$
Linearity:	$a_1 x_1(n) + a_2 x_2(n)$	$a_1 X_1(k) + a_2 X_2(k)$
Time reversal	$x(N - n)$	$X(N - k)$
Circular time shift:	$x((n - l))_N$	$X(k) e^{-j2\pi kl/N}$
Circular frequency shift:	$x(n) e^{j2\pi ln/N}$	$X((k - l))_N$
Complex conjugate:	$x^*(n)$	$X^*(N - k)$
Circular convolution:	$x_1(n) \otimes x_2(n)$	$X_1(k) X_2(k)$
Multiplication:	$x_1(n) x_2(n)$	$\frac{1}{N} X_1(k) \otimes X_2(k)$
Parseval's theorem:	$\sum_{n=0}^{N-1} x(n) y^*(n)$	$\frac{1}{N} \sum_{k=0}^{N-1} X(k) Y^*(k)$