

## The DFT

Up to this point in the course, we've learned how to analyze systems using the DTFT, defined as

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

However, the DTFT is *impossible* to use in any actual system, since  $\omega$  is a continuous variable. Therefore, infinite memory is required to store the DTFT, which is obviously infeasible. So we do the next best thing; we *sample* the DTFT in the hopes that we can maintain most of its characteristics and properties. This sampling is known as the DFT, or Discrete Fourier Transform, and represents a mapping from discrete time to *discrete* frequency.

The formal definition of the DFT is given below.

$$X[k] = \sum_{n=0}^{N-1} x[n]e^{-j\frac{2\pi kn}{N}}$$

There are a few important things to note in this definition.

- The DFT is only defined for *finite-length* sequences (those with length  $N$ ). This makes sense, as infinite-length sequences cannot be stored in a computer.
- The DFT *samples* the DTFT every  $\omega = \frac{2\pi k}{N}$ , from  $\omega = 0$  to  $\omega = 2\pi$ . That is,

$$X[k] = X_d\left(\frac{2\pi k}{N}\right)$$

Therefore, the signal length will change the number of DTFT samples that are taken. Note the different range of the DFT samples; we still get a full period, but it's shifted by  $\pi$ .

- The DFT treats both the input and the output as *periodic* with period  $N$ .

## DFT properties

Most of the properties closely mirror those of the DTFT. No DFT table should be necessary, since you can always derive a DFT by taking the DTFT and substituting  $\omega = \frac{2\pi k}{N}$ .

| DFT Property  | DT Signal                    | DFT                            |
|---------------|------------------------------|--------------------------------|
| Linearity     | $ax[n] + by[n]$              | $aX[k] + bY[k]$                |
| Time Shift    | $x[\langle n - d \rangle_N]$ | $X[k]e^{-j\frac{2\pi kd}{N}}$  |
| Modulation    | $x[n]e^{j\frac{2\pi ln}{N}}$ | $X[\langle k - l \rangle_N]$   |
| Time-Reversal | $x[\langle -n \rangle_N]$    | $X[\langle N - k \rangle_N]$   |
| Conjugation   | $x^*[n]$                     | $X^*[\langle N - k \rangle_N]$ |

In addition, there are some properties that are satisfied when  $x[n]$  is real valued:

- $X[k] = X^*[\langle N - k \rangle_N]$
- $\text{Re}\{X[k]\} = \text{Re}\{X[\langle N - k \rangle_N]\}$
- $\text{Im}\{X[k]\} = -\text{Im}\{X[\langle N - k \rangle_N]\}$
- $|X[k]| = |X[\langle N - k \rangle_N]|$
- $\angle X[k] = -\angle X[\langle N - k \rangle_N]$

## Modulo Arithmetic

Notice that most of the properties have a rather strange form; what does  $x[\langle n - d \rangle_N]$  mean? Because the DFT treats everything as periodic, we never really "shift" a signal; when we shift to the left, or to the right, the *other copies* of the signal come in. Therefore, all arithmetic performed with respect to the DFT is *modulo* arithmetic with respect to  $N$ .

The modulo operator is simple; it just returns the remainder with respect to the base. For example, if we're performing arithmetic with respect to 8, then the maximum representable number is 7, since  $7 \bmod 8 = 7$  and  $8 \bmod 8 = 0$ . We effectively "loop" back around to the beginning. This means that shifting a signal modulo  $N$  corresponds to *circularly* shifting the signal, instead of adding zeros. For example, suppose  $x[n] = \{1, 2, 3, 4, 5\}$ , and we wanted to compute  $x[\langle n - 2 \rangle_5]$ . This means that we need to *circularly shift* everything in  $x[n]$  to the right by 2; what would "pop off" of the output gets wrapped back around to the input. Therefore:

$$\begin{aligned}x[\langle n - 1 \rangle_5] &= \{5, 1, 2, 3, 4\} \\x[\langle n - 2 \rangle_5] &= \{4, 5, 1, 2, 3\}\end{aligned}$$

## Using the DFT - Circular Convolution and Zero-Padding

### Convolutions

When using the DFT, it is no longer the case that multiplication in one domain corresponds to convolution in the other domain. Instead, multiplication corresponds to *circular* convolution, since every signal is periodic. Performing circular convolution is very similar to performing linear convolution using the table method; we leave one signal untouched, we flip the other signal, and shift it to the right, but there's one major difference; the signal we flip is treated as *periodic*, so the output will also be periodic. This requires both input signals to be the *same length*. For example, suppose we have  $x[n] = \{1, 2, 3\}$  and  $\{h[n]\} = \{2, 3, 4\}$ . If we wanted to perform circular convolution, we would leave  $x[n]$  untouched, flip  $h[n]$ , but treat it as periodic:

$$\begin{array}{rcl} & 1 & 2 & 3 \\ 4 & 3 & 2 & 4 & 3 \rightarrow y[0] = 2 + 8 + 9 = 19 \\ 2 & 4 & 3 & 2 & 4 \rightarrow y[1] = 3 + 4 + 12 = 19 \\ 3 & 2 & 4 & 3 & 2 \rightarrow y[2] = 4 + 6 + 6 = 16 \end{array}$$

We stop here, since we've covered one full period of  $h[n]$ . Therefore,  $y[n] = \{19, 19, 16\}$ .

For the linear algebra-inclined, circular convolution can also be represented in matrix form. If  $x[n]$  and  $h[n]$  both have length  $N$ , then the circular convolution matrix  $H$  will be  $N \times N$ . The first column will contain  $h[n]$ , and each concurrent column is obtained by circularly shifting the previous column down by 1. In this example, we would have

$$H = \begin{bmatrix} 2 & 4 & 3 \\ 3 & 2 & 4 \\ 4 & 3 & 2 \end{bmatrix}$$

Then, the circular convolution can be performed by setting  $y = Hx$ , where  $x$  is a vector containing the input:

$$y = \begin{bmatrix} 2 & 4 & 3 \\ 3 & 2 & 4 \\ 4 & 3 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = \begin{bmatrix} 19 \\ 19 \\ 16 \end{bmatrix}$$

### Zero-Padding

We saw that, since the DFT samples the DTFT every  $\omega = \frac{2\pi k}{N}$ , increasing the length of the signal will increase the number of samples we take, therefore increasing the accuracy of the approximation. The simplest way to do so is to *zero-pad* the signal, which means adding zeros onto the *end* of the signal. This will not change the DTFT, since we're just adding zeros to the summation. However, this will improve the quality of the DFT. For example, if  $x[n] = \{1, 2, 3\}$ , taking the DFT directly only gives us 3 samples of the DTFT. However, if we zero pad, setting  $y[n] = \{1, 2, 3, 0, 0, 0, 0, 0\}$ , taking the DFT will give eight samples of the DTFT, a large improvement in quality. It's important to know how the sampled frequencies change upon zero-padding. Since  $X[k]$  and  $Y[k]$  both sample the same DTFT, we have

$$Y[k] = X_d\left(\frac{2\pi k}{8}\right) \quad \text{and} \quad X[k] = X_d\left(\frac{2\pi k}{3}\right)$$

## Practice Problems and Solutions

### Problem 1

Given that the DFT of  $x[n] = \{2, 0, 6, 4\}$  is  $\{X_0, X_1, X_2, X_3\}$ , determine the DFT of  $y[n] = \{2, 1, 0, 3\}$  and express the result in terms of  $X_0, X_1, X_2, X_3$ .

### Solution

Any time you're given a similar problem (where you want to express one DFT in terms of another DFT), you should immediately start thinking about applying some properties. First, we notice that  $\frac{1}{2}x[n] = \{1, 0, 3, 2\}$ , which is almost the desired sequence; the only problem is that it's in the wrong order. But we know how to fix this; we can *circularly shift* the signal to the right by 1 and end up with  $y[n]$ . Therefore,

$$y[n] = \frac{1}{2}x[\langle\langle n-1 \rangle\rangle_4]$$

Computing the DFT requires the linearity and time-shift properties. Since  $ax[n] \leftrightarrow aX[k]$ , and  $x[\langle\langle n-d \rangle\rangle_N] \leftrightarrow X[k]e^{-j\frac{2\pi kd}{N}}$ , we get

$$Y[k] = \frac{1}{2}e^{-j\frac{2\pi k}{4}}X[k]$$

Now, we can examine the effect on each element:

$$Y[0] = \frac{1}{2}X[0] = \frac{1}{2}X_0$$

$$Y[1] = \frac{1}{2}e^{-j\frac{\pi}{2}}X[1] = -\frac{j}{2}X_1$$

$$Y[2] = \frac{1}{2}e^{-j\pi}X[2] = -\frac{1}{2}X_2$$

$$Y[3] = \frac{1}{2}e^{-j\frac{3\pi}{2}}X[3] = \frac{j}{2}X_3$$

which tells us that

$$Y[k] = \left\{ \frac{1}{2}X_0, -\frac{j}{2}X_1, -\frac{1}{2}X_2, \frac{j}{2}X_3 \right\}$$

## Problem 2

A 3 second segment of  $x_a(t) = \cos(0.2\pi t)$  is sampled at a rate of  $T = \frac{1}{30}s$ . The resulting 90 samples are zero-padded to 128, and the DFT  $\{X[k]\}_{k=0}^{127}$  is computed. Determine the  $k_0$  that maximizes  $X[k]$  over  $0 \leq k \leq 63$ .

When we sample, we use the relation  $\omega = \Omega T$ , so the corresponding discrete-time signal is  $x[n] = \cos(\frac{\pi}{150}n)$ , from  $n = 0$  to  $n = 89$ . If  $x[n]$  were to be infinite-length, the DTFT would just be two delta functions. However, because we have to *truncate* the signal somewhere, we can think of the received signal as

$$x[n] = \cos(\frac{\pi}{150}n)w[n]$$

where  $w[n]$  is a *windowing* function defined as  $u[n] - u[n - 90]$ . Since we're multiplying two signals together in the time domain, it corresponds to a convolution in the frequency domain. We know the DTFT of  $w[n]$  - it corresponds to a discrete-time/Dirichlet sinc function. So,  $X_d(\omega)$  will be two sinc functions located at  $\frac{\pi}{150}$  and  $-\frac{\pi}{150}$ ; most importantly, the peaks will remain at  $\frac{\pi}{150}$  and  $-\frac{\pi}{150}$ .

Note the restriction on  $k$  - this is because the DTFT is  $2\pi$ -periodic, so we would also get a delta function at  $\frac{299\pi}{150}$  - there are two possible maxima. The restriction tells us that we only need to worry about the first peak.

Therefore, the problem reduces to finding the  $k$  that gets us closest to the peak, since this will maximize the value of the DFT. Since we're told the signal is zero-padded to length 128, the DFT will sample the DTFT at  $\omega = \frac{\pi k}{64}$ . Solving gives:

$$\frac{\pi k}{64} = \frac{\pi}{150} \rightarrow k = \frac{64}{150}$$

However, this is impossible, since  $k$  is a discrete variable! Therefore, we need to round to the *nearest integer*, which results in

$$\boxed{k_0 = 0}$$

### Problem 3

Determine the DFT of

$$x[n] = \cos\left(\frac{\pi n}{8}\right), 0 \leq n \leq 15$$

This problem exhibits the windowing effect; because we only take 16 samples of the cosine, we won't just get two delta functions in the DTFT. As usual, if you're asked to compute a DFT, first compute the DTFT, and then just set  $\omega = \frac{2\pi k}{N}$ . Doing so using the definition gives:

$$\begin{aligned} X_d(\omega) &= \frac{1}{2} \sum_{n=0}^{15} e^{-j(\omega - \frac{\pi}{8})n} + \frac{1}{2} \sum_{n=0}^{15} e^{-j(\omega + \frac{\pi}{8})n} \\ &= \frac{1}{2} \sum_{n=0}^{15} (e^{-j(\omega - \frac{\pi}{8})})^n + \frac{1}{2} \sum_{n=0}^{15} (e^{-j(\omega + \frac{\pi}{8})})^n \\ &= \frac{1}{2} \left[ \frac{1 - e^{-j16(\omega - \frac{\pi}{8})}}{1 - e^{-j(\omega - \frac{\pi}{8})}} \right] + \frac{1}{2} \left[ \frac{1 - e^{-j16(\omega + \frac{\pi}{8})}}{1 - e^{-j(\omega + \frac{\pi}{8})}} \right] \end{aligned}$$

Splitting the phase in both numerator and denominator gives

$$\begin{aligned} X_d(\omega) &= \frac{1}{2} \frac{e^{-j8(\omega - \frac{\pi}{8})}}{e^{-j0.5(\omega - \frac{\pi}{8})}} \frac{\sin(8(\omega - \frac{\pi}{8}))}{\sin(0.5(\omega - \frac{\pi}{8}))} + \frac{1}{2} \frac{e^{-j8(\omega + \frac{\pi}{8})}}{e^{-j0.5(\omega + \frac{\pi}{8})}} \frac{\sin(8(\omega + \frac{\pi}{8}))}{\sin(0.5(\omega + \frac{\pi}{8}))} \\ &= \frac{1}{2} e^{-j7.5(\omega - \frac{\pi}{8})} \frac{\sin(8(\omega - \frac{\pi}{8}))}{\sin(0.5(\omega - \frac{\pi}{8}))} + \frac{1}{2} e^{-j7.5(\omega + \frac{\pi}{8})} \frac{\sin(8(\omega + \frac{\pi}{8}))}{\sin(0.5(\omega + \frac{\pi}{8}))} \end{aligned}$$

This is exactly what we would expect to get. Since the DTFT of  $u[n] - u[n - 15]$  is a Dirichlet sinc, and the DTFT of  $\cos(\frac{\pi}{8}n)$  is just two delta functions at  $\omega = \frac{\pi}{8}$  and  $\omega = -\frac{\pi}{8}$ , the result is two Dirichlet sines centered at  $\omega = \frac{\pi}{8}$  and  $\omega = -\frac{\pi}{8}$ . Finally, to take the DFT, we just replace  $\omega = \frac{2\pi k}{16} = \frac{\pi k}{8}$  to end up with the result.

$$X[k] = \frac{1}{2} e^{-j7.5(\frac{\pi k}{8} - \frac{\pi}{8})} \frac{\sin(8(\frac{\pi k}{8} - \frac{\pi}{8}))}{\sin(0.5(\frac{\pi k}{8} - \frac{\pi}{8}))} + \frac{1}{2} e^{-j7.5(\frac{\pi k}{8} + \frac{\pi}{8})} \frac{\sin(8(\frac{\pi k}{8} + \frac{\pi}{8}))}{\sin(0.5(\frac{\pi k}{8} + \frac{\pi}{8}))}, k = 0, 1, \dots, 15$$

### Problem 4

Let  $X[k]$  be the 6-point DFT of  $x[n] = \{1, 2, 3, 4, 5, 6\}$ . Determine the sequence  $y[n]$  whose DFT is given by  $Y[k] = X[\langle\langle -k \rangle\rangle_6]$ .

This problem is really just a test of your understanding of modular arithmetic. We're given the time-reversal property of the DFT, which states that

$$x[\langle\langle -n \rangle\rangle_N] \leftrightarrow X[\langle\langle N - k \rangle\rangle_N]$$

However, what is  $N - k \bmod N$ ? Since the arithmetic is being performed with respect to  $N$ , we can add or subtract as many factors of  $N$  as we want without changing the result. Therefore, we can write  $N - k \bmod N$  as  $-k \bmod N$ , and  $-n \bmod N$  as  $N - n \bmod N$ . This gets everything into the form we need to solve the problem; we rewrite the property as

$$X[\langle\langle N - n \rangle\rangle_N] \leftrightarrow X[\langle\langle -k \rangle\rangle_N]$$

making it obvious that

$$y[n] = x[\langle\langle 6 - n \rangle\rangle_6]$$

To calculate this, it's simplest just to plug in values of  $n$ :

$$\begin{aligned} y[0] &= x[\langle\langle 6 \rangle\rangle_6] = x[0] \\ y[1] &= x[\langle\langle 5 \rangle\rangle_6] = x[5] \\ y[2] &= x[\langle\langle 4 \rangle\rangle_6] = x[4] \\ &\vdots \\ y[5] &= x[\langle\langle 1 \rangle\rangle_6] = x[1] \end{aligned}$$

We see that the first element remains unchanged, and all of the other elements are flipped. Therefore,

$$\boxed{y[n] = \{1, 6, 5, 4, 3, 2\}}$$