

Understanding the DFT

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

N input samples $\rightarrow N$ equally spaced frequency samples

$$X[k] = \sum_{n=0}^{N-1} x[n] \left[\cos\left(\frac{2\pi kn}{N}\right) - j \sin\left(\frac{2\pi kn}{N}\right) \right]$$

Remember: $x(t) \big|_{nTs} = x(nTs) = x[n]$

$$X[k] = \sum_{n=0}^{N-1} x[n] \left[\cos\left(\frac{2\pi kn}{N}\right) - j \sin\left(\frac{2\pi kn}{N}\right) \right]$$

Example: $N=4$ $X[k] = \sum_{n=0}^3 x[n] \left[\cos\left(\frac{2\pi kn}{4}\right) - j \sin\left(\frac{2\pi kn}{4}\right) \right]$

$$\begin{aligned} k=0: \quad X[0] &= x[0] \cos(2\pi \cdot 0 \cdot 0 \cdot 1/4) - j x[0] \sin(2\pi \cdot 0 \cdot 0 \cdot 1/4) \\ &\quad + x[1] \cos(2\pi \cdot 0 \cdot 1 \cdot 1/4) - j x[1] \sin(2\pi \cdot 0 \cdot 1 \cdot 1/4) \\ &\quad + x[2] \cos(2\pi \cdot 0 \cdot 2 \cdot 1/4) - j x[2] \sin(2\pi \cdot 0 \cdot 2 \cdot 1/4) \\ &\quad + x[3] \cos(2\pi \cdot 0 \cdot 3 \cdot 1/4) - j x[3] \sin(2\pi \cdot 0 \cdot 3 \cdot 1/4) \end{aligned}$$

$$\begin{aligned} k=1: \quad X[1] &= x[0] \cos(2\pi \cdot 1 \cdot 0 \cdot 1/4) - j x[0] \sin(2\pi \cdot 1 \cdot 0 \cdot 1/4) \\ &\quad + x[1] \cos(2\pi \cdot 1 \cdot 1 \cdot 1/4) - j x[1] \sin(2\pi \cdot 1 \cdot 1 \cdot 1/4) \\ &\quad + x[2] \cos(2\pi \cdot 1 \cdot 2 \cdot 1/4) - j x[2] \sin(2\pi \cdot 1 \cdot 2 \cdot 1/4) \\ &\quad + x[3] \cos(2\pi \cdot 1 \cdot 3 \cdot 1/4) - j x[3] \sin(2\pi \cdot 1 \cdot 3 \cdot 1/4) \end{aligned}$$

$$k=2: \quad \dots \text{etc, etc}$$

Note: Exact frequencies of different sinusoids depends on f_s and N .

$$X[k] = X(e^{j\omega}) \Big|_{\omega = \frac{2\pi k}{N}} \Rightarrow \omega = \frac{\Omega}{f_s}$$

\uparrow sample freq.

so, $\Omega = \frac{2\pi k f_s}{N} = 2\pi f_{\text{analysis}}$

$$\Rightarrow f_{\text{analysis}} = \frac{k f_s}{N}$$

Example: If $f_s = 500 \frac{\text{samples}}{\text{sec}}$, $N = 16$, $k = 1$

$$f_{\text{analysis}} = \frac{500}{16} = 31.25 \text{ Hz}$$

$$f_0 = 0 \cdot 31.25 \text{ Hz} = 0$$

$$f_1 = 1 \cdot 31.25 \text{ Hz} = 31.25 \text{ Hz}$$

$$f_2 = 2 \cdot 31.25 \text{ Hz} = 62.5 \text{ Hz}$$

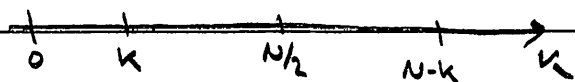
$$f_3 = 3 \cdot 31.25 \text{ Hz} = 93.75 \text{ Hz}$$

so, $X[0] \xrightarrow{\text{spectre}} \overset{\text{phase}}{\text{magnitude}} \text{ of "DC" component}$
 $X[1] \xrightarrow{\text{phase}} \text{magnitude of } 31.25 \text{ Hz component}$
 etc, etc,

Note: DFT accepts complex & real inputs

When input is real - Complex output for $k=1$ to $k=N/2-1$ are redundant with outputs for $k > N/2$

$$X[k] = X[N-k]$$



$$X[k] = |X[k]| e^{j\phi[k]} = |X[N-k]| e^{-j\phi[N-k]}$$

phase angles are opposite in sign

So, when $x[n]$ is real-valued $X[k] = X^*[N-k]$

\Rightarrow real part is even

\Rightarrow odd part is odd

• DFT Magnitudes

Real-valued sinusoid input of amp. A_0 w/ integral number of cycles over N input samples -

$$\text{DFT output magnitude} = A_0 N/2$$

Important if using fixed-point hardware.

Complex-valued sinusoid of magnitude A_0 (i.e., $A_0 e^{j2\pi f t}$) with integral number of cycles over N input samples -

$$\text{DFT output magnitude} = A_0 N$$

often see.

$$X[k] = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-j2\pi n k/N}$$

Q: How do we increase the resolution (ability to separate components at closely spaced frequencies) of the DFT output?

A: Make f_{analysis} smaller - $f_{\text{analysis}} = \frac{k f_s}{N}$

Make N larger (i.e., use more data)!

Note: Zero-padding does not improve resolution, it improves precision (via interpolation).

Rule of Thumb - To realize F_{res} Hz resolution, you must collect $\frac{1}{F_{\text{res}}}$ seconds of non-zero time samples.

• DFT processing gain - detecting signal energy embedded in noise.

- DFT output bin has $\sin(x)$ amplitude response.
- Can think of DFT output^x bin as bandpass filter with center freq. of $\frac{k f_s}{N}$
- Output mag. of DFT increases as N increases.
- DFT output bin main-lobe decreases as N increases.

Demo -

• Fast Fourier Transform

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$$\text{DFT} - X[k] = \sum_{n=0}^{N-1} x[n] e^{-j2\pi \frac{kn}{N}}$$

Q: How many mults and adds?

A: N^2 mults, N adds

FFT is a fast method for computing DFT

$$\text{FFT} \propto \frac{N \cdot \log_2 N}{2} \text{ mults}$$

Example: $N = 512$

DFT requires $(512)^2 = 262,144$

FFT requires $\approx 13,107$

20x difference

$N = 8192 \Rightarrow 1000 \times$ difference.

$N = 2,097,152 \Rightarrow$ If FFT took 10 seconds
DFT takes 3 weeks.