

DSP First, 2/e

Lecture 17

DFT: Discrete Fourier Transform

READING ASSIGNMENTS

- This Lecture:
 - Chapter 8, Sections 8-1, 8-2 and 8-4

Aug 2016

© 2003-2016, JH McClellan & RW Schaffer

3

LECTURE OBJECTIVES

- Discrete Fourier Transform

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

- DFT from DTFT by **frequency sampling**
- DFT computation (FFT)
- DFT pairs and properties
 - Periodicity in DFT (time & frequency)

Aug 2016

© 2003-2016, JH McClellan & RW Schaffer

4

Sample the DTFT → DFT

- Want **computable** Fourier transform
 - Finite signal length (L)
 - Finite number of frequencies

$$X(e^{j\hat{\omega}}) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\hat{\omega}n} \quad \rightarrow \quad X(e^{j\hat{\omega}_k}) = \sum_{n=0}^{L-1} x[n] e^{-j\hat{\omega}_k n}$$
$$\hat{\omega}_k = (2\pi/N)k, \quad k = 0, 1, 2, \dots, N-1$$

k is the frequency index

$$\text{Periodic : } X(e^{j(\hat{\omega}+2\pi)}) = X(e^{j\hat{\omega}}) \Rightarrow X[k+N] = X[k]$$

Aug 2016

© 2003-2016, JH McClellan & RW Schaffer

5

Want a Computable INVERSE Fourier Transform

- Write the inverse DTFT as a finite Riemann sum:

$$x[n] = \lim_{N \rightarrow \infty} \sum_{k=0}^{N-1} \left(X(e^{j\hat{\omega}_k}) \frac{\Delta\hat{\omega}}{2\pi} \right) e^{j(2\pi k/N)n}$$

- Note that $\left(\frac{\Delta\hat{\omega}}{2\pi}\right) = \left(\frac{2\pi/N}{2\pi}\right) = \frac{1}{N}$

- Propose:

$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j(2\pi k/N)n}, \quad \text{where } X[k] = X(e^{j\hat{\omega}_k})$$

- This is the inverse Discrete Fourier Transform (IDFT)

$$\Rightarrow X[k] = \sum_{n=0}^{N-1} x[n] e^{-j(2\pi k/N)n}, \quad \text{will be forward DFT}$$

Aug 2016

6

Inverse DFT when L=N (proof)

- Complex exponentials are ORTHOGONAL

$$\begin{aligned} x[n] &= \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j(2\pi/N)kn} \\ &= \frac{1}{N} \sum_{k=0}^{N-1} \left(\sum_{m=0}^{N-1} x[m] e^{-j(2\pi/N)km} \right) e^{j(2\pi/N)kn} \\ &= \frac{1}{N} \sum_{m=0}^{N-1} x[m] \left(\sum_{k=0}^{N-1} e^{-j(2\pi/N)km} e^{j(2\pi/N)kn} \right) \\ &= \frac{1}{N} \sum_{m=0}^{N-1} x[m] \left(\sum_{k=0}^{N-1} e^{j(2\pi/N)k(n-m)} \right) = x[n] \end{aligned}$$

$= N\delta[n-m] = \begin{cases} N & n=m \\ 0 & n \neq m \end{cases}$

Aug 2016

© 2003-2016, JH McClellan & RW Schaffer

7

Orthogonality of Complex Exponentials

The sequence set: $\left\{ e^{j(2\pi/N)kn} \right\}_{k=0}^{N-1}$ for $n = 0, 1, \dots, N-1$

$$\begin{aligned} \frac{1}{N} \sum_{n=0}^{N-1} e^{j(2\pi/N)kn} e^{-j(2\pi/N)mn} &= \frac{1}{N} \sum_{n=0}^{N-1} e^{j(2\pi/N)n(k-m)} \\ &= \frac{1}{N} \frac{1 - e^{j2\pi(k-m)}}{1 - e^{j(2\pi/N)(k-m)}} = \begin{cases} 1, & k=m \\ 0, & \text{otherwise} \end{cases} \end{aligned}$$

because $0 \leq k, m < N \quad \downarrow \quad |k-m| < N$

and $\lim_{l \rightarrow 0} \frac{1 - e^{j2\pi l}}{1 - e^{j(2\pi/N)l}} = N$

Aug 2016

© 2003-2016, JH McClellan & RW Schaffer

8

4-pt DFT: Numerical Example

- Take the 4-pt DFT of the following signal

$$x[n] = \delta[n] + \delta[n-1] \quad \{x[n]\} = [1, 1, 0, 0]$$

$$X[0] = x[0]e^{-j0} + x[1]e^{-j0} + x[2]e^{-j0} + x[3]e^{-j0} = 1+1+0+0 = 2$$

$$\begin{aligned} X[1] &= x[0]e^{-j0} + x[1]e^{-j\pi/2} + x[2]e^{-j2\pi/2} + x[3]e^{-j3\pi/2} \\ &= 1 - j = \sqrt{2}e^{-j\pi/4} \end{aligned}$$

$$X[2] = x[0]e^{-j0} + x[1]e^{-j\pi} + x[2]e^{-j2\pi} + x[3]e^{-j3\pi} = 1-1+0+0 = 0$$

$$\begin{aligned} X[3] &= x[0]e^{-j0} + x[1]e^{-j3\pi/2} + x[2]e^{-j3\pi} + x[3]e^{-j9\pi/2} \\ &= 1 + j = \sqrt{2}e^{j\pi/4} \end{aligned}$$

Aug 2016

© 2003-2016, JH McClellan & RW Schaffer

9

N-pt DFT: Numerical Example

- Take the N-pt DFT of the impulse

$$x[n] = \delta[n] \quad \{x[n]\} = [1, 0, 0, \dots, 0]$$

$$X[k] = \sum_{n=0}^{N-1} \delta[n] e^{-j(2\pi/N)kn} = \sum_{n=0}^0 \delta[n] e^{-j(2\pi/N)kn} = 1$$

$$\{X[k]\} = [1, 1, 1, \dots, 1]$$

4-pt IDFT: Numerical Example

Example 66-8: Short-Length IDFT

The 4-point DFT in Example 66-7 is $X[k] = \{2, \sqrt{2}e^{-j\pi/4}, 0, \sqrt{2}e^{j\pi/4}\}$. If we compute the 4-point IDFT of the sequence $X[k]$, we should recover $x[n]$ when we apply the IDFT summation (66.52) for each value of $n = 0, 1, 2, 3$. As before, the exponents in (66.52) will all be integer multiples of $\pi/2$ when $N = 4$.

$$x[0] = \frac{1}{4} (X[0]e^{j0} + X[1]e^{j0} + X[2]e^{j0} + X[3]e^{j0}) = \frac{1}{4} (2 + \sqrt{2}e^{-j\pi/4} + 0 + \sqrt{2}e^{j\pi/4}) = 1$$

$$x[1] = \frac{1}{4} (X[0]e^{j\pi/2} + X[1]e^{j\pi/2} + X[2]e^{j\pi} + X[3]e^{j3\pi/2}) = \frac{1}{4} (2 + \sqrt{2}e^{j(-\pi/4+\pi/2)} + 0 + \sqrt{2}e^{j(\pi/4+3\pi/2)}) = \frac{1}{4} (2 + (1+j) + (1-j)) = 1$$

$$x[2] = \frac{1}{4} (X[0]e^{j\pi} + X[1]e^{j\pi} + X[2]e^{j2\pi} + X[3]e^{j3\pi}) = \frac{1}{4} (2 + \sqrt{2}e^{j(-\pi/4+\pi)} + 0 + \sqrt{2}e^{j(\pi/4+3\pi)}) = \frac{1}{4} (2 + (-1+j) + (-1-j)) = 0$$

$$x[3] = \frac{1}{4} (X[0]e^{j3\pi/2} + X[1]e^{j3\pi/2} + X[2]e^{j3\pi} + X[3]e^{j9\pi/2}) = \frac{1}{4} (2 + \sqrt{2}e^{j(-\pi/4+3\pi/2)} + 0 + \sqrt{2}e^{j(\pi/4+9\pi/2)}) = \frac{1}{4} (2 + (-1-j) + (-1+j)) = 0$$

Thus we recover the signal $x[n] = \{1, 1, 0, 0\}$ from its DFT coefficients, $X[k] = \{2, \sqrt{2}e^{-j\pi/4}, 0, \sqrt{2}e^{j\pi/4}\}$.

Matrix Form for N-pt DFT

- In MATLAB, NxN DFT matrix is `dfmtx(N)`
 - Obtain DFT by $\mathbf{X} = \text{dfmtx}(N) * \mathbf{x}$
 - Or, **more efficiently by** $\mathbf{X} = \text{fft}(\mathbf{x}, N)$
 - Fast Fourier transform (FFT) algorithm later

$$\begin{bmatrix} X[0] \\ X[1] \\ X[2] \\ \vdots \\ X[N-1] \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & e^{-j2\pi/N} & e^{-j4\pi/N} & \dots & e^{-j2(N-1)\pi/N} \\ 1 & e^{-j4\pi/N} & e^{-j8\pi/N} & \dots & e^{-j4(N-1)\pi/N} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & e^{-j2(N-1)\pi/N} & e^{-j4(N-1)\pi/N} & \dots & e^{-j2(N-1)(N-1)\pi/N} \end{bmatrix}}_{\text{DFT matrix}} \begin{bmatrix} x[0] \\ x[1] \\ x[2] \\ \vdots \\ x[N-1] \end{bmatrix}$$

Signal vector

FFT: Fast Fourier Transform

- FFT is an **algorithm** for computing the DFT
- $N \log_2 N$ versus N^2 operations
 - Count multiplications (and additions)
 - For example, when $N = 1024 = 2^{10}$
 - $\approx 10,000$ ops vs. $\approx 1,000,000$ operations
 - ≈ 1000 times faster
- What about $N=256$, how much faster?

Zero-Padding gives denser FREQUENCY SAMPLING

- Want many samples of DTFT
 - WHY? to make a smooth plot
 - Finite signal length (L)
 - Finite number of frequencies (N)
 - Thus, we need $L < N, N \rightarrow \infty, X[k] \rightarrow X(e^{j\hat{\omega}})$

$$X(e^{j\hat{\omega}_k}) = \sum_{n=0}^{L-1} x[n] e^{-j\hat{\omega}_k n}$$

$$\hat{\omega}_k = (2\pi/N)k, \quad k = 0, 1, 2, \dots, N-1$$

Zero-Padding with the FFT

- Get many samples of DTFT
 - Finite signal length (L)
 - Finite number of frequencies (N)
 - Thus, we need $L < N, N \rightarrow \infty, X[k] \rightarrow X(e^{j\hat{\omega}})$

In MATLAB

- Use `x = fft(x,N)`
- With `L=length(x)` less than N
- Define `xpadtoN = [x,zeros(1,N-L)];`
- Take the N-pt DFT of `xpadtoN`

DFT periodic in k (frequency domain)

- Since DTFT is periodic in frequency, the DFT must also be periodic in k

$$X[k] = X(e^{j(2\pi/N)k})$$

$$X[k + N] = X(e^{j(2\pi/N)(k+N)}) = X(e^{j(2\pi/N)k + j(2\pi/N)N}) = X(e^{j(2\pi/N)k})$$

- What about Negative indices and Conjugate Symmetry?

$$X(e^{-j(2\pi/N)k}) = X^*(e^{j(2\pi/N)k})$$

$$\Rightarrow X[-k] = X^*[k]$$

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

$$N = 32 \Rightarrow$$

$$X[31] = X^*[1]$$

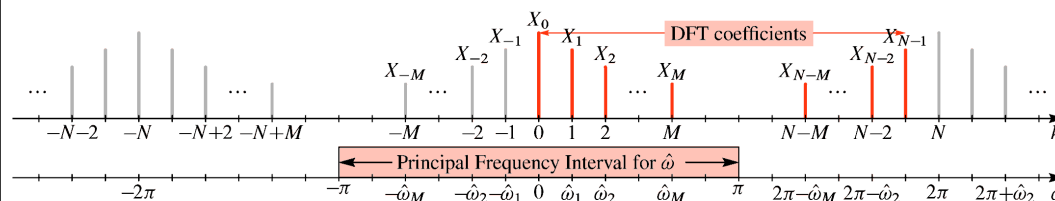
$$X[30] = X^*[2]$$

$$X[29] = X^*[3]$$

DFT Periodicity in Frequency Index

$$X[k] = X(e^{j\hat{\omega}_k}) = X(e^{j(2\pi/N)k})$$

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



$$X[k + N] = X[k] \Leftrightarrow X(e^{j(\hat{\omega} + 2\pi)}) = X(e^{j\hat{\omega}})$$

$$\Rightarrow X[N - k] = X[-k],$$

e.g., $X[N - 2] = X[-2]$

DFT pairs & properties

- Recall DTFT pairs because DFT is sampled DTFT
 - See next two slides
- DFT acts on a finite-length signal, so we can use DTFT pairs & properties for finite signals
- Want DFT properties related to computation
- And, we will concentrate on one more pair:
 - DTFT and DFT of finite sinusoid (or cexp)
 - Length-L signal
 - N-pt DFT

Table of DTFT Pairs

Time-Domain: $x[n]$	Frequency-Domain: $X(e^{j\hat{\omega}})$
$\delta[n]$	1
$\delta[n - n_d]$	$e^{-j\hat{\omega}n_d}$
$u[n] - u[n - L]$	$\frac{\sin(\frac{1}{2}L\hat{\omega})}{\sin(\frac{1}{2}\hat{\omega})} e^{-j\hat{\omega}(L-1)/2}$
$\frac{\sin(\hat{\omega}_b n)}{\pi n}$	$\frac{1}{2} [u(\hat{\omega} + \hat{\omega}_b) + u(\hat{\omega} - \hat{\omega}_b)] = \begin{cases} 1 & \hat{\omega} \leq \hat{\omega}_b \\ 0 & \hat{\omega}_b < \hat{\omega} \leq \pi \end{cases}$
$a^n u[n] \quad (a < 1)$	$\frac{1}{1 - ae^{-j\hat{\omega}}}$
$b^n u[-n] \quad (b > 1)$	$\frac{1}{1 - be^{-j\hat{\omega}}}$

These 3 signals have infinite length

Table 8-1 Basic discrete Fourier transform pairs.

Table of DFT Pairs	
Time-Domain: $x[n]$	Frequency-Domain: $X[k]$
$\delta[n]$	1
$\delta[n - n_d]$	$e^{-j(2\pi k/N)n_d}$
$r_L[n] = u[n] - u[n - L]$	$\frac{\sin(\frac{1}{2}L(2\pi k/N))}{\sin(\frac{1}{2}(2\pi k/N))} e^{-j(2\pi k/N)(L-1)/2}$ $= D_L(2\pi k/N)$
$r_L[n] e^{j(2\pi k_0/N)n}$	$D_L(2\pi(k - k_0)/N) e^{-j(2\pi(k-k_0)/N)(L-1)/2}$

Table of DTFT Properties

Property Name	Time-Domain: $x[n]$	Frequency-Domain: $X(e^{j\hat{\omega}})$
Periodic in $\hat{\omega}$		$X(e^{j(\hat{\omega}+2\pi)}) = X(e^{j\hat{\omega}})$
Linearity	$ax_1[n] + bx_2[n]$	$aX_1(e^{j\hat{\omega}}) + bX_2(e^{j\hat{\omega}})$
Conjugate Symmetry	$x[n]$ is real	$X(e^{-j\hat{\omega}}) = X^*(e^{j\hat{\omega}})$
Conjugation	$x^*[n]$	$X^*(e^{-j\hat{\omega}})$
Time-Reversal	$x[-n]$	$X(e^{-j\hat{\omega}})$
Delay	$x[n - n_d]$	$e^{-j\hat{\omega}n_d} X(e^{j\hat{\omega}})$
Frequency Shift	$x[n] e^{j\hat{\omega}_0 n}$	$X(e^{j(\hat{\omega}-\hat{\omega}_0)})$
Modulation	$x[n] \cos(\hat{\omega}_0 n)$	$\frac{1}{2} X(e^{j(\hat{\omega}-\hat{\omega}_0)}) + \frac{1}{2} X(e^{j(\hat{\omega}+\hat{\omega}_0)})$
Convolution	$x[n] * h[n]$	$X(e^{j\hat{\omega}}) H(e^{j\hat{\omega}})$
Parseval's Theorem	$\sum_{n=-\infty}^{\infty} x[n] ^2$	$\frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\hat{\omega}}) ^2 d\hat{\omega}$

These 3 properties involve circular indexing

Table of DFT Properties		
Property Name	Time-Domain: $x[n]$	Frequency-Domain: $X[k]$
Periodic	$x[n] = x[n + N]$	$X[k] = X[k + N]$
Linearity	$ax_1[n] + bx_2[n]$	$aX_1[k] + bX_2[k]$
Conjugate Symmetry	$x[n]$ is real	$X[N - k] = X^*[k]$
Conjugation	$x^*[n]$	$X^*[N - k]$
Time-Reversal	$x[((N - n))_N]$	$X[N - k]$
Delay	$x[((n - n_d))_N]$	$e^{-j(2\pi k/N)n_d} X[k]$
Frequency Shift	$x[n]e^{j(2\pi k_0/N)n}$	$X[k - k_0]$
Modulation	$x[n] \cos((2\pi k_0/N)n)$	$\frac{1}{2}X[k - k_0] + \frac{1}{2}X[k + k_0]$
Convolution	$\sum_{m=0}^{N-1} h[m]x[((n - m))_N]$	$H[k]X[k]$
Parseval's Theorem	$\sum_{n=0}^{N-1} x[n] ^2 = \frac{1}{N} \sum_{k=0}^{N-1} X[k] ^2$	

Convolution Property not the same

- Almost true for DFT:
 - Convolution maps to multiplication of transforms
- Need a different kind of convolution
 - CIRCULAR CONVOLUTION
 - LATER in an advanced DSP course
- Likewise, for Time-Shifting
 - Has to be circular
 - Because the “n” domain is also periodic

Delay Property of DFT

- Recall DTFT property for time shifting:

$$y[n] = x[n - n_d] \Leftrightarrow Y(e^{j\hat{\omega}}) = X(e^{j\hat{\omega}})e^{-j\hat{\omega}n_d}$$
- Expected DFT property via **frequency sampling**

$$y[n] = x[\underbrace{n - n_d}] \Leftrightarrow Y[k] = X[k]e^{-j(2\pi k/N)n_d}$$
- Indices such as $n - n_d$ must be evaluated modulo-N because

$$e^{-j(2\pi k/N)(n_d+N)} = e^{-j(2\pi k/N)n_d}$$

DTFT of a Length-L Pulse

- Know DTFT of finite rectangular pulse
 - Dirichlet form and a linear phase term

$$x[n] = \begin{cases} 1 & 0 \leq n < L \\ 0 & \text{otherwise} \end{cases} \Leftrightarrow X(e^{j\hat{\omega}}) = \frac{\sin(\frac{1}{2}L\hat{\omega})}{\sin(\frac{1}{2}\hat{\omega})} e^{-j\hat{\omega}(L-1)/2}$$

$$D_L(\hat{\omega}) = \frac{\sin(\frac{1}{2}L\hat{\omega})}{\sin(\frac{1}{2}\hat{\omega})}$$

- Use frequency-sampling to get DFT

$$X[k] = \frac{\sin(\frac{1}{2}L(2\pi k/N))}{\sin(\frac{1}{2}(2\pi k/N))} e^{-j(2\pi k/N)(L-1)/2} = D_L(2\pi k/N) e^{-j(\pi k/N)(L-1)}$$

DTFT of a Finite Length Complex Exponential (1)

- Know DTFT of finite rectangular pulse
 - Dirichlet form and a linear phase term

$$x[n] = \begin{cases} 1 & 0 \leq n < L \\ 0 & \text{otherwise} \end{cases} \Leftrightarrow X(e^{j\hat{\omega}}) = \underbrace{\frac{\sin(\frac{1}{2}L\hat{\omega})}{\sin(\frac{1}{2}\hat{\omega})}}_{D_L(\hat{\omega})} e^{-j\hat{\omega}(L-1)/2}$$

$$D_L(\hat{\omega}) = \frac{\sin(\frac{1}{2}L\hat{\omega})}{\sin(\frac{1}{2}\hat{\omega})}$$

- Use frequency-shift property

$$y[n] = \begin{cases} e^{j\hat{\omega}_0 n} & 0 \leq n < L \\ 0 & \text{otherwise} \end{cases} \Leftrightarrow Y(e^{j\hat{\omega}}) = \frac{\sin(\frac{1}{2}L(\hat{\omega} - \hat{\omega}_0))}{\sin(\frac{1}{2}(\hat{\omega} - \hat{\omega}_0))} e^{-j(\hat{\omega} - \hat{\omega}_0)(L-1)/2}$$

Aug 2016

© 2003-2016, JH McClellan & RW Schaefer

26

DTFT of a Finite Length Complex Exponential (2)

- Know DTFT, so we can sample in frequency

$$y[n] = \begin{cases} e^{j\hat{\omega}_0 n} & 0 \leq n < L \\ 0 & \text{otherwise} \end{cases} \Leftrightarrow Y(e^{j\hat{\omega}}) = \frac{\sin(\frac{1}{2}L(\hat{\omega} - \hat{\omega}_0))}{\sin(\frac{1}{2}(\hat{\omega} - \hat{\omega}_0))} e^{-j(\hat{\omega} - \hat{\omega}_0)(L-1)/2}$$

- Thus, the N-point DFT is

$$y[n] = \begin{cases} e^{j\hat{\omega}_0 n} & 0 \leq n < L \\ 0 & L \leq n < N \end{cases} \Leftrightarrow Y[k] = Y(e^{j\hat{\omega}}) \text{ at } \hat{\omega} = \frac{2\pi k}{N}$$

$$Y[k] = \frac{\sin(\frac{1}{2}L(\frac{2\pi k}{N} - \hat{\omega}_0))}{\sin(\frac{1}{2}(\frac{2\pi k}{N} - \hat{\omega}_0))} e^{-j(\frac{2\pi k}{N} - \hat{\omega}_0)(L-1)/2}$$

Dirichlet Function

$$D_L(\hat{\omega}) = \frac{\sin(\frac{1}{2}L\hat{\omega})}{\sin(\frac{1}{2}\hat{\omega})}$$

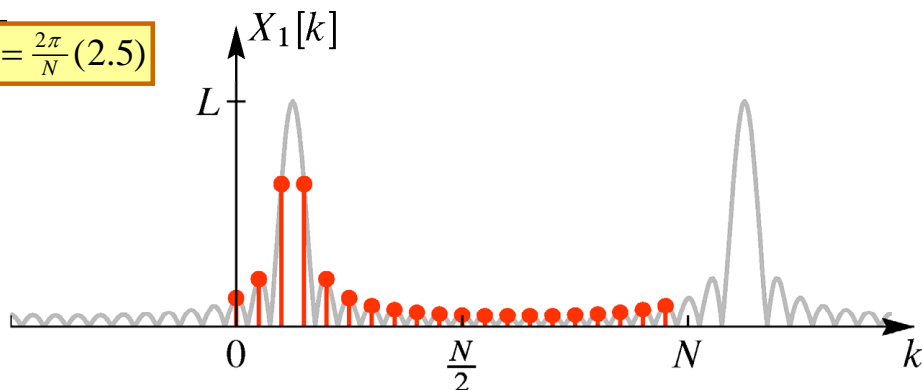
Aug 2016

© 2003-2016, JH McClellan & RW Schaefer

27

20-pt DFT of Complex Exponential

$$\hat{\omega}_0 = \frac{2\pi}{N}(2.5)$$



$$x_1[n] = \begin{cases} e^{j\hat{\omega}_0 n} & 0 \leq n < L \\ 0 & \text{otherwise} \end{cases} \Leftrightarrow X_1[k] = \frac{\sin(\frac{1}{2}L(\frac{2\pi k}{N} - \hat{\omega}_0))}{\sin(\frac{1}{2}(\frac{2\pi k}{N} - \hat{\omega}_0))} e^{-j(\frac{2\pi k}{N} - \hat{\omega}_0)(L-1)/2}$$

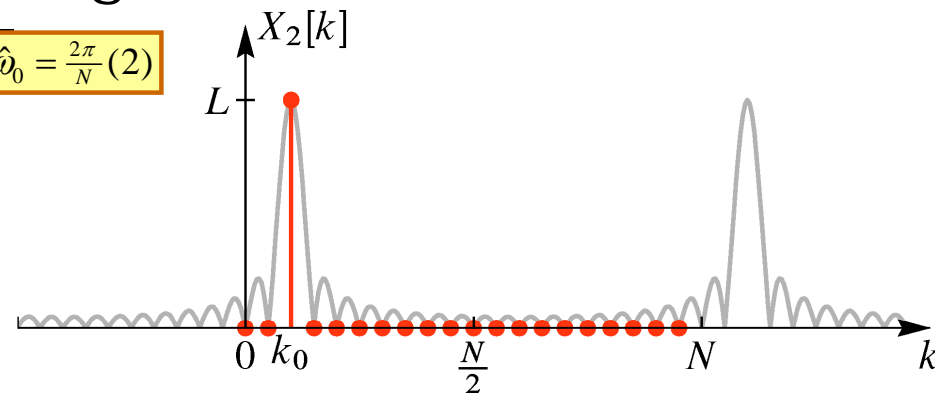
$$D_L(\hat{\omega}) = \frac{\sin(\frac{1}{2}L\hat{\omega})}{\sin(\frac{1}{2}\hat{\omega})} \text{ so the outline is a shifted Dirichlet form}$$

Aug 2016

28

20-pt DFT of Complex Exp: "on the grid"

$$\hat{\omega}_0 = \frac{2\pi}{N}(2)$$



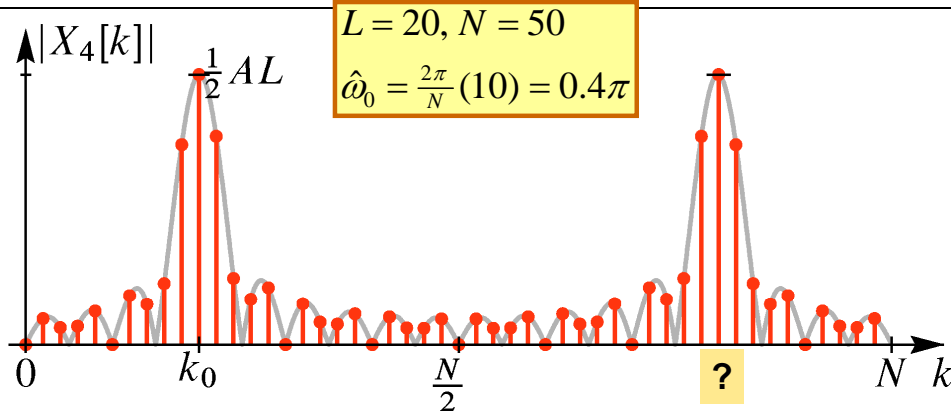
$$x_2[n] = \begin{cases} e^{j\hat{\omega}_0 n} & 0 \leq n < L \\ 0 & \text{otherwise} \end{cases} \Leftrightarrow X_2[k] = \frac{\sin(\frac{1}{2}L(\frac{2\pi k}{N} - \hat{\omega}_0))}{\sin(\frac{1}{2}(\frac{2\pi k}{N} - \hat{\omega}_0))} e^{-j(\frac{2\pi k}{N} - \hat{\omega}_0)(L-1)/2}$$

$$D_L(\hat{\omega}) = \frac{\sin(\frac{1}{2}L\hat{\omega})}{\sin(\frac{1}{2}\hat{\omega})} \text{ so the outline is a shifted Dirichlet form}$$

Aug 2016

29

50-pt DFT of Sinusoid: zero padding



$$L = 20, N = 50$$

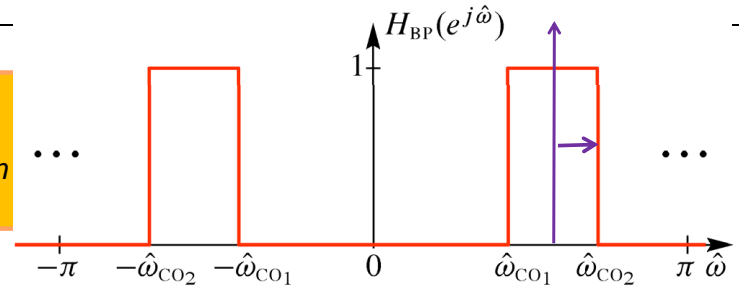
$$\hat{\omega}_0 = \frac{2\pi}{N} (10) = 0.4\pi$$

$$x_4[n] = \begin{cases} A \cos(\hat{\omega}_0 n) & 0 \leq n < L \\ 0 & \text{otherwise} \end{cases} \Leftrightarrow$$

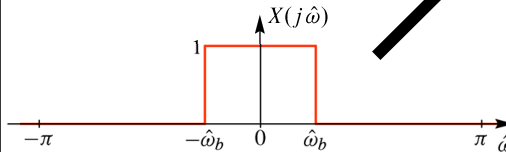
$$X_4[k] = \frac{1}{2} AD_L \left(\frac{2\pi k}{N} - \hat{\omega}_0 \right) e^{-j \left(\frac{2\pi k}{N} - \hat{\omega}_0 \right) (L-1)/2} + \frac{1}{2} AD_L \left(\frac{2\pi k}{N} + \hat{\omega}_0 \right) e^{-j \left(\frac{2\pi k}{N} + \hat{\omega}_0 \right) (L-1)/2}$$

RECALL: BandPass Filter (BPF)

Frequency shifting up and down is done by cosine multiplication in the time domain



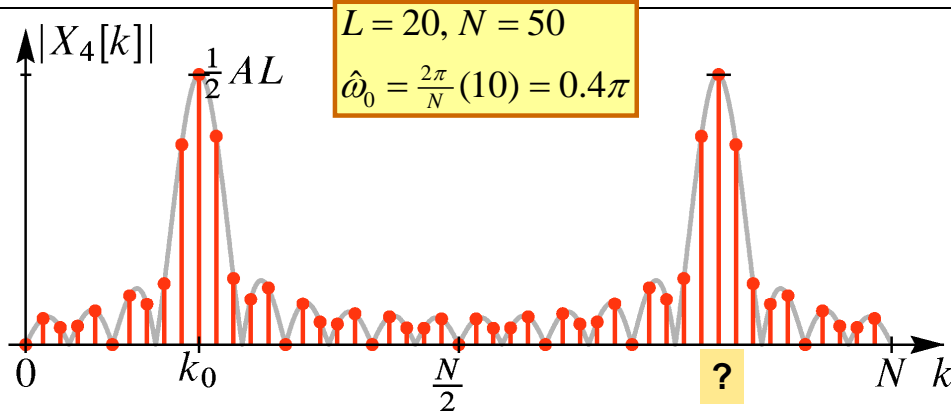
BPF is frequency shifted version of LPF (below)



$$h_{BP}[n] = 2 \cos(\hat{\omega}_{mid} n) \frac{\sin(\frac{1}{2} \hat{\omega}_{diff} n)}{\pi n}$$

$$\Leftrightarrow H_{BP}(e^{j\hat{\omega}}) = \begin{cases} 0 & |\hat{\omega}| \leq \hat{\omega}_{co1} \\ 1 & \hat{\omega}_{co1} < |\hat{\omega}| \leq \hat{\omega}_{co2} \\ 0 & \hat{\omega}_{co2} < |\hat{\omega}| \leq \pi \end{cases}$$

50-pt DFT of Sinusoid: zero padding



$$L = 20, N = 50$$

$$\hat{\omega}_0 = \frac{2\pi}{N} (10) = 0.4\pi$$

Zero-crossings of Dirichlet?
Width of Dirichlet?
Density of frequency samples?

Thus we have a simple BPF