

Lecture 08:

Discrete Fourier Transform

Outline

- Discrete Fourier Transform
- Convolution with the DFT
- Short-Time Fourier Transform

Fourier Transforms

	<i>Time</i>	<i>Frequency</i>
Fourier Series (FS)	Continuous periodic $\tilde{x}(t)$	Discrete infinite c_k
Fourier Transform (FT)	Continuous infinite $x(t)$	Continuous infinite $X(\Omega)$
Discrete-Time FT (DTFT)	Discrete infinite $x[n]$	Continuous periodic $X(e^{j\omega})$
Discrete FT (DFT)	Discrete finite/pdc $\tilde{x}[n]$	Discrete finite/pdc $X[k]$

Discrete FT (DFT)

Discrete FT (DFT)	Discrete finite/pdc $x[n]$	Discrete finite/pdc $X[k]$
----------------------	-------------------------------	-------------------------------

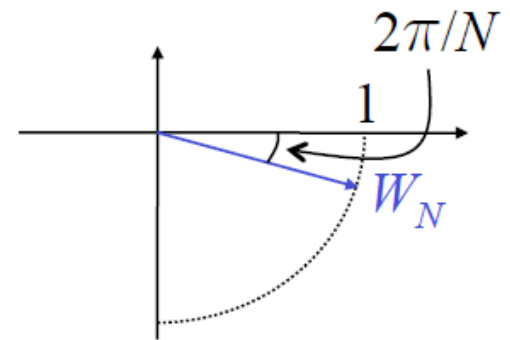
- A *finite* or *periodic* sequence has only N unique values, $x[n]$ for $0 \leq n < N$
- Spectrum is completely defined by N distinct frequency samples
- Divide $0..2\pi$ into N equal steps,
$$\{\omega_k\} = \frac{2\pi k}{N}$$

DFT and IDFT

- Uniform sampling of DTFT spectrum:

$$X[k] = X(e^{j\omega}) \Big|_{\omega=\frac{2\pi k}{N}} = \sum_{n=0}^{N-1} x[n] e^{-j\frac{2\pi k}{N}n}$$

- DFT:** $X[k] = \sum_{n=0}^{N-1} x[n] W_N^{kn}$



where $W_N = e^{-j\frac{2\pi}{N}}$ i.e. $-1/N^{\text{th}}$ of a revolution

IDFT

- Inverse DFT: **IDFT** $x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] W_N^{-nk}$
- Check:

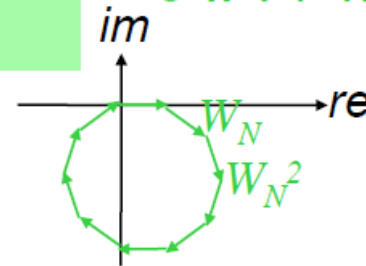
$$x[n] = \frac{1}{N} \sum_k \left(\sum_l x[l] W_N^{kl} \right) W_N^{-nk}$$

$$= \frac{1}{N} \sum_{l=0}^{N-1} x[l] \sum_{k=0}^{N-1} W_N^{k(l-n)}$$

Sum of complete set
of rotated vectors
= 0 if $l \neq n$; = N if $l = n$

$$= x[n]$$

$$0 \leq n < N$$



or finite
geometric series
 $= (1 - W_N^{lN}) / (1 - W_N^l)$

DFT examples

- **Finite impulse** $x[n] = \begin{cases} 1 & n = 0 \\ 0 & n = 1 \dots N - 1 \end{cases}$

$$\Rightarrow X[k] = \sum_{n=0}^{N-1} x[n] W_N^{kn} = W_N^0 = 1 \quad \forall k$$

- **Periodic sinusoid:**

$$x[n] = \cos\left(\frac{2\pi rn}{N}\right) \quad (r \in \mathbb{Z}) = \frac{1}{2} (W_N^{-rn} + W_N^{rn})$$

$$\Rightarrow X[k] = \frac{1}{2} \sum_{n=0}^{N-1} (W_N^{-rn} + W_N^{rn}) W_N^{kn}$$

$$(0 \leq k < N) = \begin{cases} \frac{N}{2} & k = r, k = N - r \\ 0 & \text{otherwise} \end{cases}$$

DFT: Matrix form

- $X[k] = \sum_{n=0}^{N-1} x[n] \cdot W_N^{kn}$ as a matrix multiply:

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

■ i.e.


$$\mathbf{X} = \mathbf{D}_N \cdot \mathbf{x}$$

Matrix IDFT

- If $\mathbf{X} = \mathbf{D}_N \cdot \mathbf{x}$
then $\mathbf{x} = \mathbf{D}_N^{-1} \cdot \mathbf{X}$
- i.e. inverse DFT is also just a matrix,

$$\mathbf{D}_N^{-1} = \frac{1}{N} \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & W_N^{-1} & W_N^{-2} & \dots & W_N^{-(N-1)} \\ 1 & W_N^{-2} & W_N^{-4} & \dots & W_N^{-2(N-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & W_N^{-(N-1)} & W_N^{-2(N-1)} & \dots & W_N^{-(N-1)^2} \end{bmatrix}$$

$$= 1/N D_N^*$$

DFT and MATLAB

- MATLAB is concerned with *sequences* not continuous functions like $X(e^{j\omega})$
- Instead, we use the DFT to sample $X(e^{j\omega})$ on an (arbitrarily-fine) grid:
 - `X = freqz(x,1,w);` samples the DTFT of sequence `x` at angular frequencies in `w`
 - `X = fft(x);` calculates the N -point DFT of an N -point sequence `x`

DFT and DTFT

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

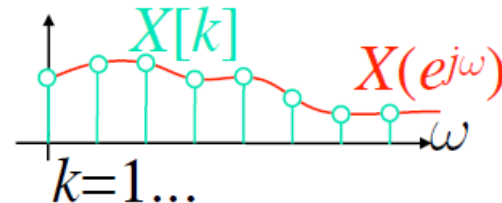
- *continuous freq ω*
- *infinite $x[n]$, $-\infty < n < \infty$*

DFT $X[k] = \sum_{n=0}^{N-1} x[n]W_N^{kn}$

- *discrete freq $k=N\omega/2\pi$*
- *finite $x[n]$, $0 \leq n < N$*

- DFT ‘samples’ DTFT at discrete freqs:

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



DTFT from DFT

- N -point DFT completely specifies the continuous DTFT of the finite sequence

$$X(e^{j\omega}) = \sum_{n=0}^{N-1} \left(\frac{1}{N} \sum_{k=0}^{N-1} X[k] W_N^{-kn} \right) e^{-j\omega n}$$

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

$$\Delta\omega_k = \omega - \frac{2\pi k}{N}$$

interpolation

$$= \frac{1}{N} \sum_{k=0}^{N-1} X[k] \cdot \frac{\sin N \frac{\Delta\omega_k}{2}}{\sin \frac{\Delta\omega_k}{2}} \cdot e^{-j \frac{(N-1)}{2} \Delta\omega_k}$$

“periodic sinc”

Periodic sinc

$$\sum_{n=0}^{N-1} e^{-j\Delta\omega_k n} = \frac{1 - e^{-jN\Delta\omega_k}}{1 - e^{-j\Delta\omega_k}}$$

*factor out
half the angle*

$$= \frac{e^{-jN\Delta\omega_k/2}}{e^{-j\Delta\omega_k/2}} \cdot \frac{e^{jN\Delta\omega_k/2} - e^{-jN\Delta\omega_k/2}}{e^{j\Delta\omega_k/2} - e^{-j\Delta\omega_k/2}}$$

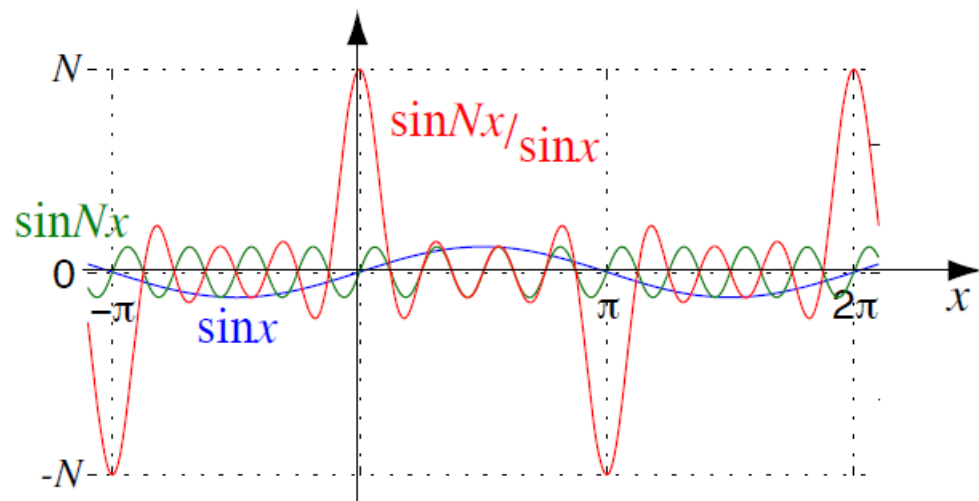
$$= e^{-j\frac{(N-1)}{2}\Delta\omega_k} \frac{\sin N \frac{\Delta\omega_k}{2}}{\sin \frac{\Delta\omega_k}{2}} \leftarrow \text{pure real}$$

pure phase \rightarrow

- $= N$ when $\Delta\omega_k = 0$; $= (-)N$ when $\Delta\omega_k/2 = \pi$
- $= 0$ when $\Delta\omega_k/2 = r \cdot \pi/N$, $r = \pm 1, \pm 2, \dots$
- other values in-between...

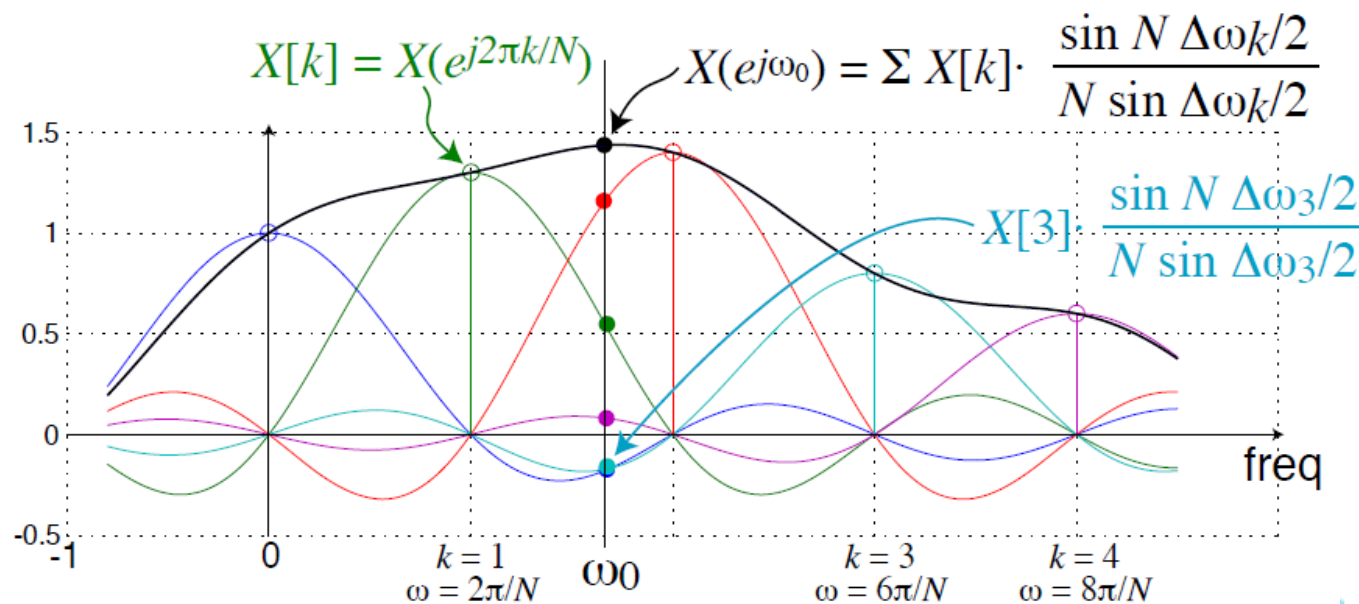
Periodic sinc

$$\frac{\sin Nx}{\sin x}$$



DFT \rightarrow DTFT
 = interpolation
 by periodic
 sinc

$$X[k] \rightarrow X(e^{j\omega})$$



DFT from overlength DTFT

- If $x[n]$ has more than N points, can still form $X[k] = X(e^{j\omega}) \Big|_{\omega = \frac{2\pi k}{N}}$
- IDFT of $X[k]$ will give N point $\tilde{x}[n]$
- How does $\tilde{x}[n]$ relate to $x[n]$?

DFT from overlength DTFT

$$\begin{array}{ccccc}
 x[n] & \xrightarrow{\text{DTFT}} & X(e^{j\omega}) & \xrightarrow{\text{sample}} & X[k] & \xrightarrow{\text{IDFT}} & \tilde{x}[n] \\
 -A \leq n < B & & & & & & 0 \leq n < N
 \end{array}$$

$$\begin{aligned}
 \tilde{x}[n] &= \frac{1}{N} \sum_{k=0}^{N-1} \left(\sum_{\ell=-\infty}^{\infty} x[\ell] W_N^{k\ell} \right) W_N^{-nk} \\
 &= \sum_{\ell=-\infty}^{\infty} x[\ell] \left(\frac{1}{N} \sum_{k=0}^{N-1} W_N^{k(\ell-n)} \right)
 \end{aligned}$$

$= 1$ for $n - \ell = rN, r \in \mathbb{I}$
 $= 0$ otherwise

$$\Rightarrow \tilde{x}[n] = \sum_{r=-\infty}^{\infty} x[n - rN]$$

all values shifted by exact multiples of N pts to lie in $0 \leq n < N$

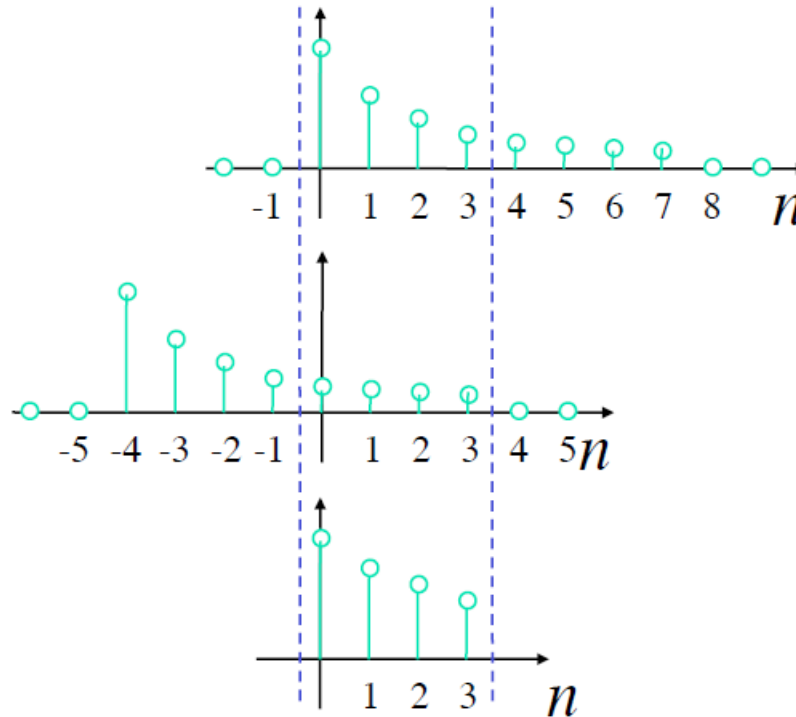
DFT from DTFT example

- If $x[n] = \{ 8, 5, 4, 3, 2, 2, 1, 1 \}$ (8 point)
- We form $X[k]$ for $k = 0, 1, 2, 3$
by sampling $X(e^{j\omega})$ at $\omega = 0, \pi/2, \pi, 3\pi/2$
- IDFT of $X[k]$ gives 4 pt $\tilde{x}[n] = \sum_{r=-\infty}^{\infty} x[n - rN]$
- Overlap only for $r = -1$: ($N = 4$)

$$\Rightarrow \tilde{x}[n] = \left\{ \begin{array}{cccc} 8 & 5 & 4 & 3 \\ + & + & + & + \\ 2 & 2 & 2 & 1 \end{array} \right\} = \{10 \quad 7 \quad 5 \quad 4\}$$

DFT from DTFT example

- $x[n]$
- $x[n+N]$
($r = -1$)
- $\tilde{x}[n]$



- $\tilde{x}[n]$ is the **time aliased** or ‘folded down’ version of $x[n]$.

Properties: Circular time shift

- DFT properties mirror DTFT, with twists:
- Time shift must stay within N -pt 'window'

$$g[\langle n - n_0 \rangle_N] \quad \longleftrightarrow \quad W_N^{kn_0} G[k]$$

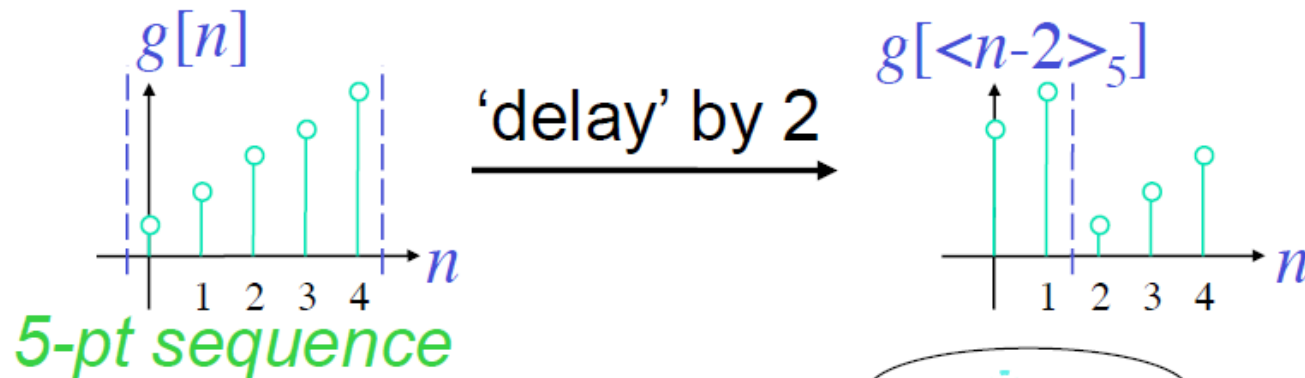
- Modulo- N indexing keeps index between 0 and $N-1$:

$$g[\langle n - n_0 \rangle_N] = \begin{cases} g[n - n_0] & n \geq n_0 \\ g[N + n - n_0] & n < n_0 \end{cases}$$

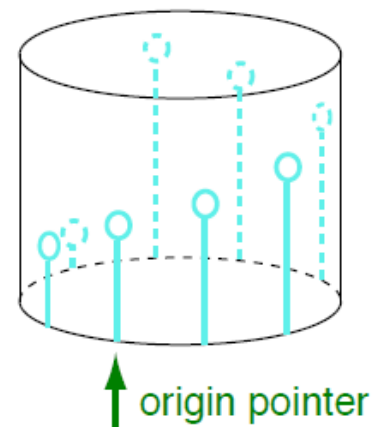
$0 \leq n_0 < N$

Circular time shift

- Points shifted out to the right don't disappear – they come in from the left



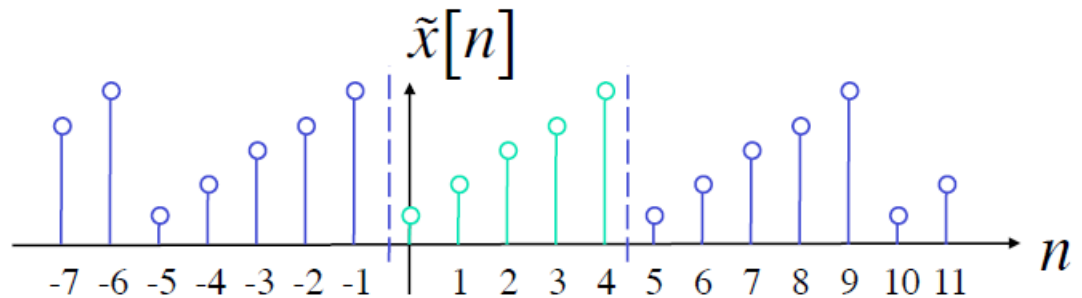
- Like a 'barrel shifter':



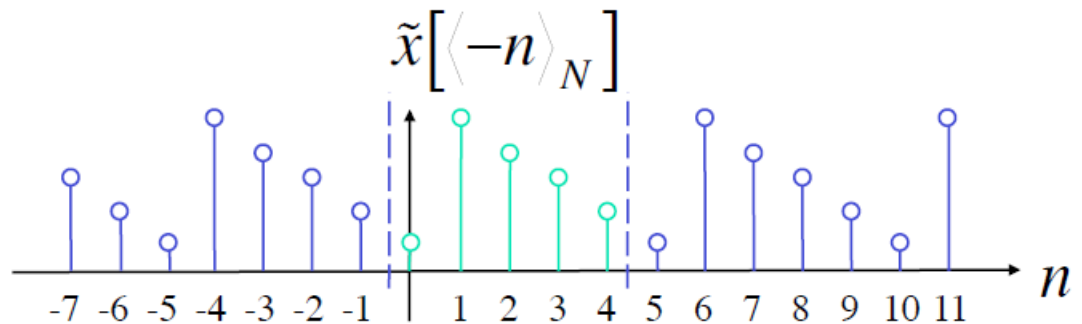
Circular time reversal

- Time reversal is tricky in 'modulo- N ' indexing - **not** reversing the sequence:

*5-pt sequence
made periodic*



*Time-reversed
periodic sequence*



- Zero point stays fixed; remainder flips

Duality

- DFT and IDFT are very similar
 - both map an N -pt vector to an N -pt vector

- Duality:

if $g[n] \leftrightarrow G[k]$

then $G[n] \leftrightarrow N \cdot g[\langle -k \rangle_N]$

Circular time reversal

- i.e. if you treat DFT sequence as a **time** sequence, result is almost symmetric

2. Convolution with the DFT

- IDTFT of product of DTFTs of two N -pt sequences is their $2N-1$ pt convolution
- IDFT of the product of two N -pt DFTs can only give N points!
- Equivalent of $2N-1$ pt result **time aliased**:
 - i.e. $y_c[n] = \sum_{r=-\infty}^{\infty} y_l[n + rN] \quad (0 \leq n < N)$
 - must be, because $G[k]H[k]$ are exact samples of $G(e^{j\omega})H(e^{j\omega})$
- This is known as **circular convolution**

Circular convolution

- Can also do entire convolution with modulo- N indexing
- Hence, **Circular Convolution:**

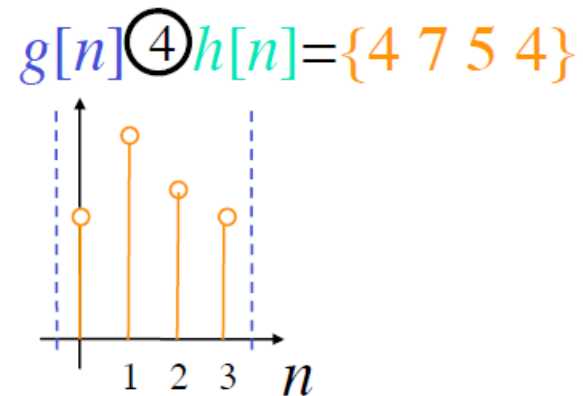
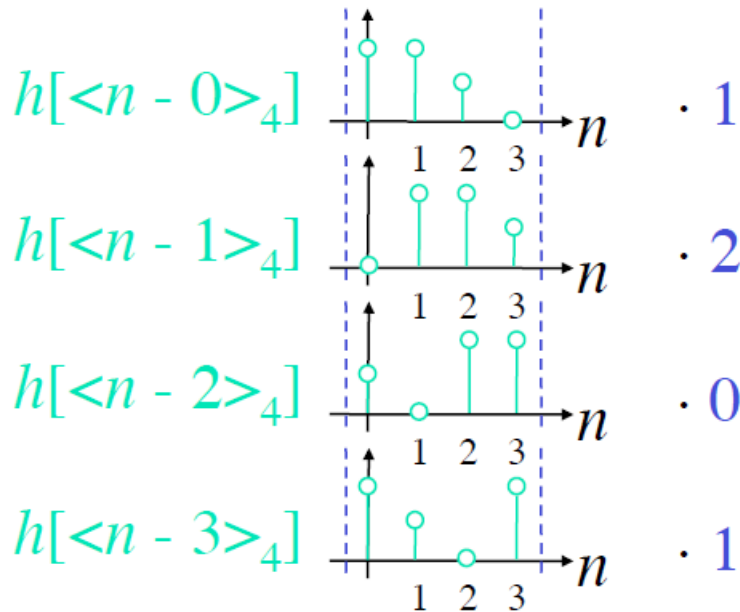
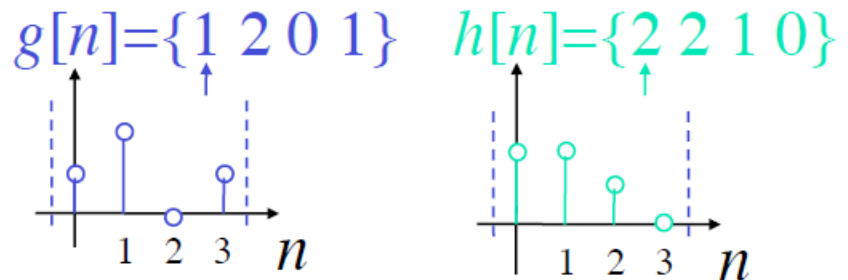
$$\sum_{m=0}^{N-1} g[m]h[\langle n-m \rangle_N] \leftrightarrow G[k]H[k]$$

- Written as $g[n] \circledcirc h[n]$

Circular convolution example

■ 4 pt sequences:

$$\sum_{m=0}^{N-1} g[m]h[\langle n-m \rangle_N]$$



check: $g[n] \circledast h[n]$
 $= \{2 \ 6 \ 5 \ 4 \ 2 \ 1 \ 0\}$

DFT properties summary

- Circular convolution

$$\sum_{m=0}^{N-1} g[m]h[\langle n - m \rangle_N] \leftrightarrow G[k]H[k]$$

- Modulation

$$g[n] \cdot h[n] \leftrightarrow \frac{1}{N} \sum_{m=0}^{N-1} G[m]H[\langle k - m \rangle_N]$$

- Duality

$$G[n] \leftrightarrow N \cdot g[\langle -k \rangle_N]$$

- Parseval

$$\sum_{n=0}^{N-1} |x[n]|^2 = \frac{1}{N} \sum_{k=0}^{N-1} |X[k]|^2$$

Linear convolution w/ the DFT

- DFT \rightarrow fast **circular** convolution
- .. but we need **linear** convolution
- Circular conv. is **time-aliased** linear conv.; can aliasing be avoided?
- e.g. convolving L -pt $g[n]$ with M -pt $h[n]$:
 $y[n] = g[n] \circledast h[n]$ has $L+M-1$ nonzero pts
- Set DFT size $N \geq L+M-1 \rightarrow$ **no aliasing**

Linear convolution w/ the DFT

- Procedure ($N = L + M - 1$):

- pad L -pt $g[n]$ with (at least) $M-1$ zeros

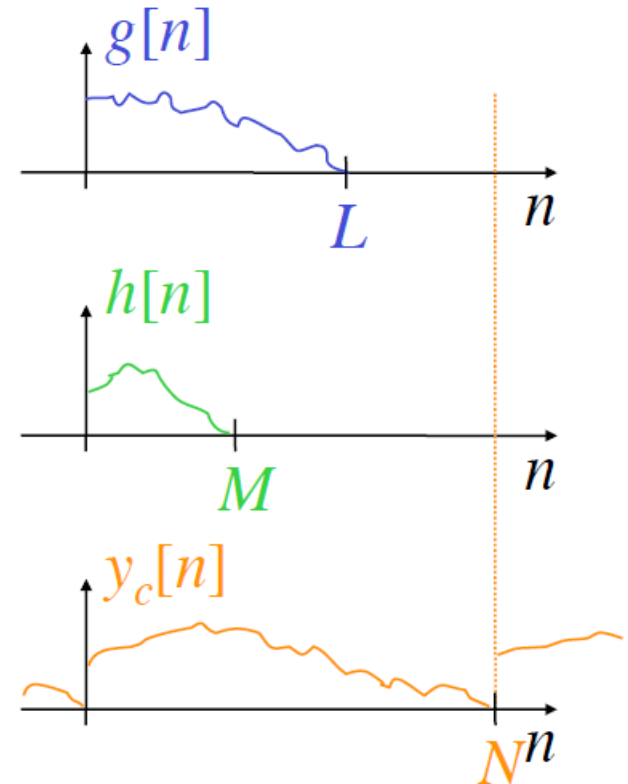
→ N -pt DFT $G[k]$, $k = 0..N-1$

- pad M -pt $h[n]$ with (at least) $L-1$ zeros

→ N -pt DFT $H[k]$, $k = 0..N-1$

- $Y[k] = G[k] \cdot H[k]$, $k = 0..N-1$

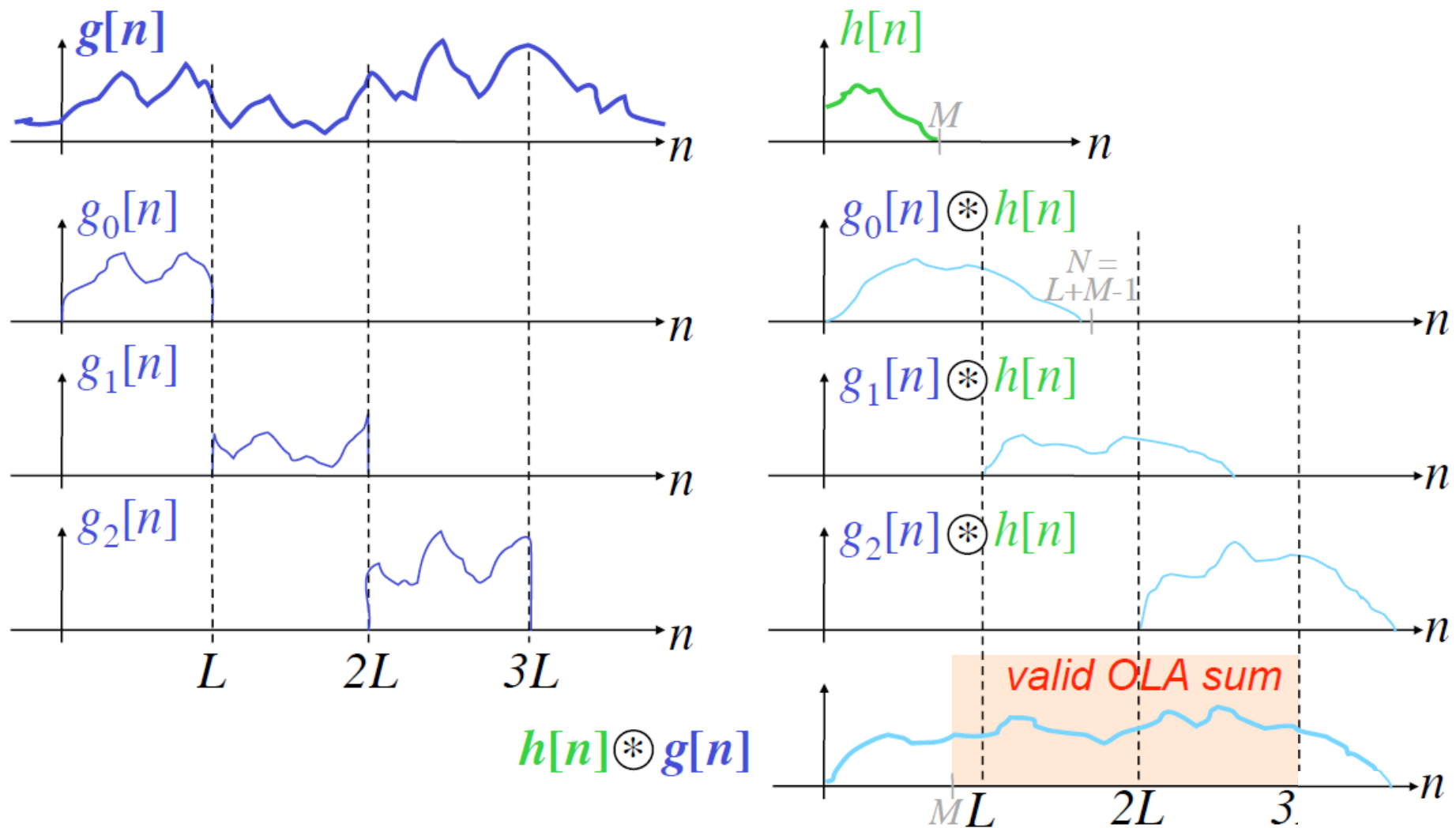
- $\text{IDFT}\{Y[k]\} = \sum_{r=-\infty}^{\infty} y_L[n + rN] = y_L[n] \quad (0 \leq n < N)$



Overlap-Add convolution

- Very long $g[n]$ → break up into segments, convolve **piecewise**, **overlap**
→ bound size of DFT, processing delay
- Make $g_i[n] = \begin{cases} g[n] & i \cdot N \leq n < (i + 1) \cdot N \\ 0 & \text{otherwise} \end{cases}$
 $\Rightarrow g[n] = \sum_i g_i[n]$
 $\Rightarrow h[n] \circledast g[n] = \sum_i h[n] \circledast g_i[n]$
- Called Overlap-Add (**OLA**) convolution

Overlap-Add convolution



DFT of Truncated Signals

- What if the signal is not time-limited?
We can think of limiting the sum to N points as a truncation of the signal:

$$x_w[n] = w[n]x[n]$$

$$w[n] = \begin{cases} 1, & n = 0, 1, 2, \dots, N \\ 0, & \text{otherwise} \end{cases}$$

- What are the implications of this in the frequency domain?
(Hint: convolution)

- Popular Windows:

- Rectangular:

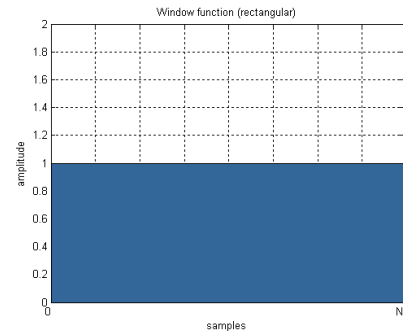
$$w[n] = \begin{cases} 1, & n = 0, 1, 2, \dots, N \\ 0, & \text{otherwise} \end{cases}$$

- Generalized Hanning:

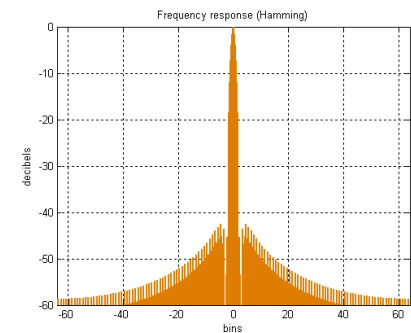
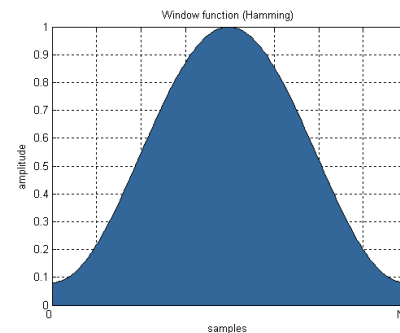
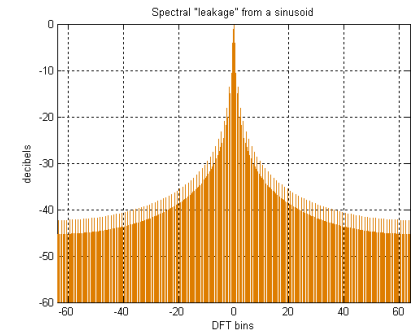
$$w[n] = \alpha + (1 - \alpha) \cos\left(\frac{2\pi n}{N-1}\right)$$

- Triangular:

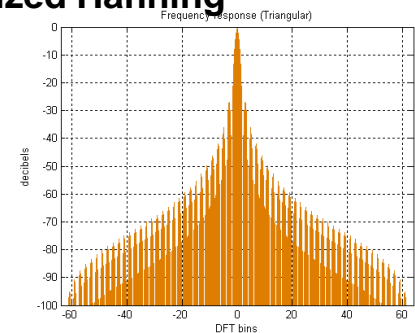
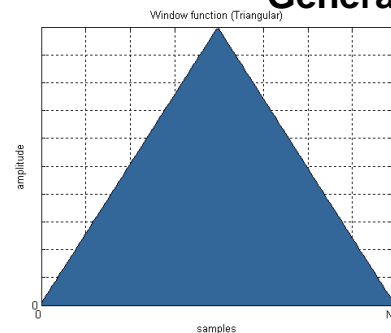
$$w[n] = \frac{2}{N} \left(\frac{N}{2} - \left| n - \frac{N-1}{2} \right| \right)$$



- Rectangular



- Generalized Hanning



- Triangular

Impact on Spectral Estimation

- The spectrum of a windowed sinewave is the convolution of two impulse functions with the frequency response of the window.
- For two closely spaced sinewaves, there is “leakage” between each sinewave’s spectrum.

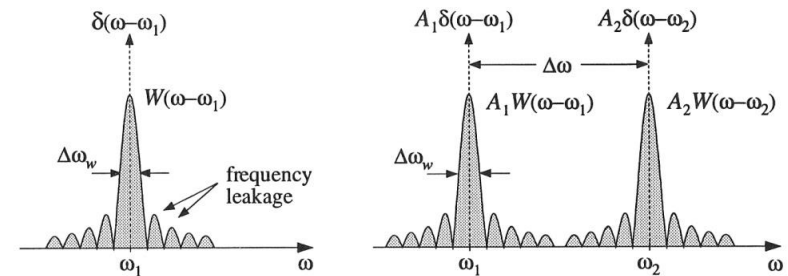


Fig. 9.1.3 Spectra of windowed single and double sinusoids.

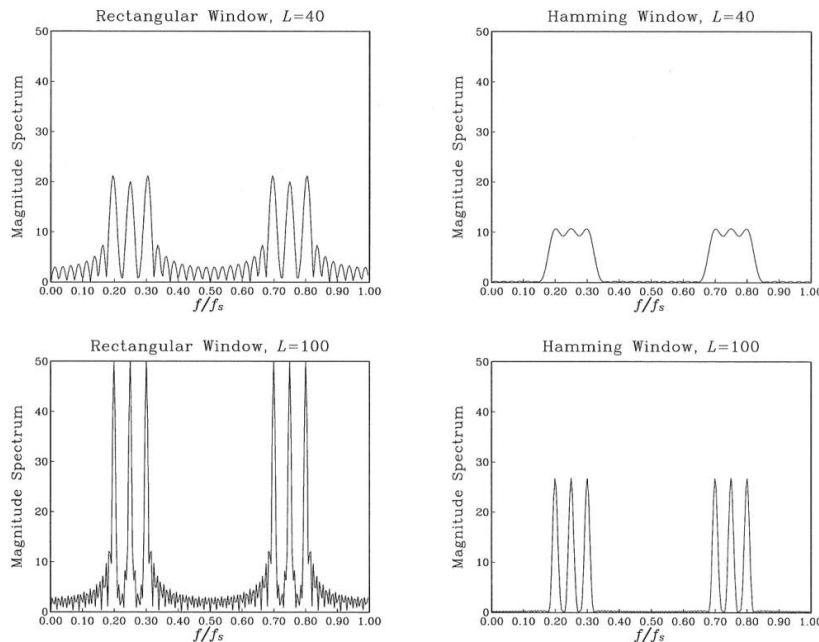


Fig. 9.1.9 Rectangular and Hamming spectra for $L = 40$ and 100 .

- The impact of this leakage can be mitigated by using a window function with a narrower main lobe.
- For example, consider the spectrum of three sinewaves computed using a rectangular and a Hamming window.
- We see that for the same number of points, the spectrum produced by the Hamming window separates the sinewaves.
- What is the computational cost?