

DFT: Discrete Fourier Transform

Congruence (Integer modulo m)

- In this section, all letters stand for integers.
- $\gcd(n, m)$ = the greatest common divisor of n and m
- Let $d = \gcd(n, m)$
All the linear combinations $r \cdot n + s \cdot m$ of n and m are multiples of d .
- $a|b \Leftrightarrow a$ is a divisor of b .
- In an expression $\text{mod } m$, m is a strictly positive integer.
- $a = b \text{ mod } m \Leftrightarrow m | b - a$
- If $a = b \text{ mod } m$
 $c = d \text{ mod } m$
then
 - $a + c = (b + d) \text{ mod } m$
 - $a - c = (b - d) \text{ mod } m$
 - $a \cdot c = (b \cdot d) \text{ mod } m$
 - $a^n = b^n \text{ mod } m$ with $n > 0$
 - $n \cdot a = (n \cdot b) \text{ mod } m$
- $\langle m \rangle_N = m \text{ mod } N = m + rN$ for unique r that make $0 \leq m + rN < N$

N-point signal

- $x[n]$: **N-point signal** \Leftrightarrow
 - Has finite duration
 - Duration interval $\subset [0, N)$
 - **Cyclic shift** of $x[n]$ by n_0 ; $0 \leq n < N$; $0 \leq n_0 < N$:

$$= x[\langle n - n_0 \rangle_N] = \begin{cases} x[n - n_0] & ; n_0 \leq n < N \\ x[N + n - n_0] & ; 0 \leq n < n_0 \\ 0 & ; \text{otherwise} \end{cases}$$

- $x[\langle n - n_0 \rangle_N]$ is another N-point signal
- Example
 - $N = 5$

n	-2	-1	0	1	2	3	4	5	6	7
$x[n]$	0	0	a	b	c	d	e	0	0	0

$$x[\langle n-2 \rangle_5] \quad | \quad 0 \quad | \quad 0 \quad | \quad d \quad | \quad e \quad | \quad a \quad | \quad E \quad | \quad F \quad | \quad | \quad |$$

- **N-point circular convolution** of N-point signal

$$x_1[n] \otimes x_2[n] = \sum_{m=0}^{N-1} x_1[m] x_2[\langle n-m \rangle_N] = \sum_{m=0}^{N-1} x_1[\langle n-m \rangle_N] x_2[m]$$

Proof

$$\begin{aligned} \sum_{m=0}^{N-1} x_1[\langle n-m \rangle_N] x_2[m] &= \sum_{m=0}^n x_1[n-m] x_2[m] + \sum_{m=n+1}^{N-1} x_1[N+n-m] x_2[m] \\ &= \sum_{\ell=0}^n x_1[\ell] x_2[n-\ell] + \sum_{\ell=n+1}^{N-1} x_1[\ell] x_2[n+N-\ell] \\ &= \sum_{\ell=0}^n x_1[\ell] x_2[\langle n-\ell \rangle_N] + \sum_{\ell=n+1}^{N-1} x_1[\ell] x_2[\langle n-\ell \rangle_N] \end{aligned}$$

- If $x_1[m]$ has many 0's, use $\sum_{m=0}^{N-1} x_1[m] x_2[\langle n-m \rangle_N]$, eliminating each of $x_2[\langle n-m \rangle_N]$ that is multiplied by $x_1[m] = 0$

- Example:

- $N = 3$

$$\text{Let } y[n] = x_1[n] \otimes x_2[n]$$

$$y[n] = \sum_{m=0}^2 x_1[m] x_2[\langle n-m \rangle_3] = x_1[0] x_2[n] + x_1[1] x_2[\langle n-1 \rangle_3] + x_1[2] x_2[\langle n-2 \rangle_3]$$

$$y[0] = x_1[0] x_2[0] + x_1[1] x_2[\langle -1 \rangle_3] + x_1[2] x_2[\langle -2 \rangle_3]$$

$$= x_1[0] x_2[0] + x_1[1] x_2[2] + x_1[2] x_2[3]$$

$$y[1] = x_1[0] x_2[1] + x_1[1] x_2[\langle 0 \rangle_3] + x_1[2] x_2[\langle -1 \rangle_3]$$

$$= x_1[0] x_2[1] + x_1[1] x_2[0] + x_1[2] x_2[2]$$

$$y[2] = x_1[0] x_2[2] + x_1[1] x_2[\langle 1 \rangle_3] + x_1[2] x_2[\langle 0 \rangle_3]$$

$$= x_1[0] x_2[2] + x_1[1] x_2[1] + x_1[2] x_2[0]$$

- To find $x_1[n] \otimes x_2[n]$ using circular convolution rule

- $\hat{X}_1[k] = x_1[0] + x_1[1] \mathbf{y}_3^{-k} + x_1[2] \mathbf{y}_3^{-2k} = x_1[0] + x_1[1] A + x_1[2] A^2$

$$\hat{X}_2[k] = x_2[0] + x_2[1] \mathbf{y}_3^{-k} + x_2[2] \mathbf{y}_3^{-2k} = x_2[0] + x_2[1] A + x_2[2] A^2$$

The multiplication $\hat{X}_1[k] \cdot \hat{X}_2[k]$ can be easily find with TI calculator or Mathcad (select A, then choose Symbolics > Polynomial Coefficients):

$$\hat{X}_1[k] \cdot \hat{X}_2[k] = \begin{bmatrix} 1 & A & A^2 & A^3 & A^4 \end{bmatrix} \begin{bmatrix} x_{10}x_{20} \\ x_{10}x_{21} + x_{11}x_{20} \\ x_{10}x_{22} + x_{11}x_{21} + x_{12}x_{20} \\ x_{11}x_{22} + x_{12}x_{21} \\ x_{12}x_{22} \end{bmatrix}$$

But $A^3 = 1, A^4 = A$; therefore,

$$\hat{X}_1[k] \cdot \hat{X}_2[k] = \begin{bmatrix} 1 & A & A^2 \end{bmatrix} \begin{bmatrix} x_{10}x_{20} + x_{11}x_{22} + x_{12}x_{21} \\ x_{10}x_{21} + x_{11}x_{20} + x_{12}x_{22} \\ x_{10}x_{22} + x_{11}x_{21} + x_{12}x_{20} \end{bmatrix}.$$

Obviously (actually from inverse DFT), $\begin{bmatrix} y[0] \\ y[1] \\ y[2] \end{bmatrix} = \begin{bmatrix} x_{10}x_{20} + x_{11}x_{22} + x_{12}x_{21} \\ x_{10}x_{21} + x_{11}x_{20} + x_{12}x_{22} \\ x_{10}x_{22} + x_{11}x_{21} + x_{12}x_{20} \end{bmatrix}.$

• Example:

• $x_1[n] = 2 \ 3 \ 1 \ 1, x_2[n] = 1 \ 0 \ 1 \ 2$

$$\begin{aligned} \hat{X}_1[k] \cdot \hat{X}_2[k] &= (2 + 3\mathbf{y}_4^{-k} + \mathbf{y}_4^{-2k} + \mathbf{y}_4^{-3k}) \cdot (1 + 0\mathbf{y}_4^{-k} + \mathbf{y}_4^{-2k} + 2\mathbf{y}_4^{-3k}) \\ &= 2 + 3\mathbf{y}_4^{-k} + 3\mathbf{y}_4^{-2k} + 8\mathbf{y}_4^{-3k} + 8\mathbf{y}_4^{-4k} + 3\mathbf{y}_4^{-5k} + 2\mathbf{y}_4^{-6k} \\ &= 10 + 6\mathbf{y}_4^{-k} + 5\mathbf{y}_4^{-2k} + 8\mathbf{y}_4^{-3k} \end{aligned}$$

$$x_1[n] \otimes x_2[n] = 10 \ 6 \ 5 \ 8$$

DFT

- $\mathbf{y}_N = e^{j\frac{2\pi}{N}}$
 - $(\mathbf{y}_N)^N = 1$
 - $\overline{\mathbf{y}_N} = (\mathbf{y}_N)^{-1}$
 - $N \bmod p = 0 \Rightarrow \mathbf{y}_N^p = \mathbf{y}_{\frac{N}{p}}$
 - $\mathbf{y}_{kN}^N = \mathbf{y}_k$
 - For even n , $\mathbf{y}_{\frac{N}{2}}^2 = e^{j\pi} = -1$
 - $\sum_{n=0}^{N-1} \mathbf{y}_N^{-kn} = \begin{cases} N & k = mN \\ 0 & \text{otherwise} \end{cases}$

$$\text{Proof } (\mathbf{y}_N)^N = \left(e^{j\frac{2\pi}{N}} \right)^N = e^{j2\pi} = 1$$

Proof $\overline{y_N} = e^{\overline{j\frac{2p}{N}}} = e^{-j\frac{2p}{N}} = \left(e^{j\frac{2p}{N}} \right)^{-1}$

Proof $N \bmod p = 0 \Rightarrow y_N^p = \left(e^{j\frac{2p}{N}} \right)^p = e^{j\frac{2p}{N/p}} = y_{\frac{N}{p}}$

Proof $\sum_{n=0}^{N-1} e^{j2pn\frac{k}{N}} = \begin{cases} N ; \text{if } \frac{k}{N} \in I \\ 0 ; \text{if } \frac{k}{N} \notin I \end{cases}$

$$\sum_{n=0}^{N-1} y_N^{-kn} = \sum_{n=0}^{N-1} e^{j\frac{2p}{N}(-kn)} = \sum_{n=0}^{N-1} e^{j\frac{2p}{N}nk} = \begin{cases} N ; \text{if } \frac{k}{N} \in I \\ 0 ; \text{if } \frac{k}{N} \notin I \end{cases}$$

• $[\Psi_N]_{pq} = y_N^{-(p-1)(q-1)} ; 1 \leq p, q \leq N$

• $\Psi_N = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & y_N^{-1} & y_N^{-2} & \dots & y_N^{-(N-1)} \\ 1 & y_N^{-2} & y_N^{-4} & \dots & y_N^{-2(N-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & y_N^{-(N-1)} & y_N^{-2(N-1)} & \dots & y_N^{-(N-1)(N-1)} \end{bmatrix}$

• $\Psi_N^{-1} = \frac{1}{N} \overline{\Psi_N}$

Proof

$$\begin{aligned} [\overline{\Psi_N} \Psi_N]_{pq} &= \sum_{\ell=1}^N [\overline{\Psi_N}]_{p\ell} [\Psi_N]_{\ell q} = \sum_{\ell=1}^N y_N^{(p-1)(\ell-1)} y_N^{-(\ell-1)(q-1)} \\ &= \sum_{\ell=1}^N y_N^{(\ell-1)(p-q)} = \sum_{m=0}^{N-1} y_N^{m(p-q)} \\ &= \begin{cases} N & ; p = q \Rightarrow y_N^{p-q} = 1 \\ \frac{1 - y_N^{N(p-q)}}{1 - y_N^{(p-q)}} = 0 & ; p \neq q \end{cases} \\ &= N I_N \end{aligned}$$

• **N-point DFT** of the N-point signal x[n]:

$$\hat{X}[k] = \hat{X}\left(\mathbf{w} = k\frac{2p}{N}\right) = \sum_{n=0}^{N-1} x[n] y_N^{-nk} ; 0 \leq k < N$$

- $\hat{X}[k] = \hat{X}\left(\mathbf{w} = k \frac{2\mathbf{p}}{N}\right) = \sum_{n=0}^{N-1} x[n] e^{-jnk \frac{2\mathbf{p}}{N}} = \sum_{n=0}^{N-1} x[n] \mathbf{y}_N^{-nk} \quad ; 0 \leq k < N$

- $$\begin{bmatrix} \hat{X}[0] \\ \hat{X}[1] \\ \hat{X}[2] \\ \vdots \\ \hat{X}[n] \end{bmatrix} = \Psi_N \begin{bmatrix} x[0] \\ x[1] \\ x[2] \\ \vdots \\ x[n] \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & \cdots & 1 \\ 1 & \mathbf{y}_N^{-1} & \mathbf{y}_N^{-2} & \cdots & \mathbf{y}_N^{-(N-1)} \\ 1 & \mathbf{y}_N^{-2} & \mathbf{y}_N^{-4} & \cdots & \mathbf{y}_N^{-2(N-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \mathbf{y}_N^{-(N-1)} & \mathbf{y}_N^{-2(N-1)} & \cdots & \mathbf{y}_N^{-(N-1)(N-1)} \end{bmatrix} \begin{bmatrix} x[0] \\ x[1] \\ x[2] \\ \vdots \\ x[n] \end{bmatrix}$$

- $\mathbf{d}[n] \xleftrightarrow{DFT} \sum_{n=0}^{N-1} \mathbf{d}[n] \mathbf{y}_N^{-nk} = \mathbf{y}_N^{-0k} = 1$

- $(DFT^{-1}) x[n] = \frac{1}{N} \sum_{k=0}^{N-1} \hat{X}[k] \mathbf{y}_N^{nk} \xleftrightarrow{DFT} \hat{X}[k] = \sum_{n=0}^{N-1} x[n] \mathbf{y}_N^{-nk} \quad (DFT)$

Proof $\underline{x} = \begin{bmatrix} x[0] \\ x[1] \\ x[2] \\ \vdots \\ x[n] \end{bmatrix} ; \underline{\hat{X}} = \begin{bmatrix} \hat{X}[0] \\ \hat{X}[1] \\ \hat{X}[2] \\ \vdots \\ \hat{X}[n] \end{bmatrix}$

$\Rightarrow \underline{\hat{X}} = \Psi_N \underline{x} \Rightarrow \underline{x} = \frac{1}{N} \Psi_N^* \underline{\hat{X}}$

- Given tool for computing DFT

$$\hat{X}[k] \xrightarrow{DFT^{-1}} \boxed{\text{DFT}} \xrightarrow{\boxed{\text{Time reverse}}} \boxed{\times \frac{1}{N}} \rightarrow x[n]$$

DFT^{-1}

Proof After DFT-finder $g[q] = \sum_{p=0}^{N-1} \hat{X}[p] \mathbf{y}_N^{-pq}$,

Time reverse: $g[N-q] = \sum_{p=0}^{N-1} \hat{X}[p] \mathbf{y}_N^{-p(N-q)} = \sum_{p=0}^{N-1} \hat{X}[p] \mathbf{y}_N^{pq}$

- $\hat{X}[k] \xrightarrow{DFT^{-1}} x[n] \xrightarrow{DFT} \hat{X}(\mathbf{w})$
- $x[\langle n - n_0 \rangle] \xleftrightarrow{DFT} \mathbf{y}_N^{-kn_0} \hat{X}[k]$

Proof

$$\begin{aligned}
\sum_{n=0}^{N-1} x[\langle n - n_0 \rangle_N] \mathbf{y}_N^{-nk} &= \sum_{n=n_0}^{N-1} x[n - n_0] \mathbf{y}_N^{-nk} + \sum_{n=0}^{n_0-1} x[N + n - n_0] \mathbf{y}_N^{-nk} \\
&= \sum_{\ell=0}^{N-n_0-1} x[\ell] \mathbf{y}_N^{-(\ell+n_0)k} + \sum_{\ell=N-n_0}^{N-1} x[\ell] \mathbf{y}_N^{-(\ell+n_0-N)k} \\
&= \sum_{\ell=0}^{N-1} x[\ell] \mathbf{y}_N^{-(\ell+n_0)k}
\end{aligned}$$

- $\mathbf{d}[n - n_0] \xleftarrow{DFT} \mathbf{y}_N^{-n_0 k} ; 0 \leq n_0 < N$

$$\text{Proof } \mathbf{d}[\langle n - n_0 \rangle_N] = \begin{cases} \mathbf{d}[n - n_0] & ; n_0 \leq n < N \\ \mathbf{d}[N + n - n_0] & ; 0 \leq n < n_0 \\ 0 & ; \text{otherwise} \end{cases} = \mathbf{d}[n - n_0]$$

$$\mathbf{d}[\langle n - n_0 \rangle_N] \xrightarrow{DFT} \mathbf{y}_N^{-kn_0} \times 1 \quad \begin{matrix} 0 \leq n < N \\ 0 \leq k < N \end{matrix}$$

- **Circular convolution rule:**

$$x_1[n] \otimes x_2[n] \xleftrightarrow{DFT} \hat{X}_1[k] \cdot \hat{X}_2[k]$$

Proof

$$\begin{aligned}
\sum_{n=0}^{N-1} \left(\sum_{m=0}^{N-1} x_1[m] x_2[\langle n - m \rangle_N] \right) \mathbf{y}_N^{-nk} &= \sum_{m=0}^{N-1} x_1[m] \left(\sum_{n=0}^{N-1} x_2[\langle n - m \rangle_N] \mathbf{y}_N^{-nk} \right) \\
&= \sum_{m=0}^{N-1} x_1[m] (\hat{X}_2[k] \mathbf{y}_N^{-mk})
\end{aligned}$$

- To use DFT to compute regular convolutions of time-limited (finite-duration) signal:

If $x_1[n]$ has duration interval $0 \leq n < N_1$

$x_2[n]$ has duration interval $0 \leq n < N_2$

Let $N = N_1 + N_2 - 1$,

then $x_1[n] * x_2[n] = x_1[n] \otimes x_2[n]$ for $0 \leq n < N$ and = 0 otherwise

To see this,

$x_1[n] * x_2[n]$ has finite duration at most $N = N_1 + N_2 - 1$

Think of $x_1[n]$ and $x_2[n]$ as N -point signal whose last values are 0 (0-padding)

- **Block convolution**

$h[n]$ has duration interval $0 \leq n < P$ (impulse response of a causal FIR system)

$w[n]$ = a (possibly) infinite-duration signal

To compute $h[n] * w[n]$

- Divide $w[n]$ into blocks of some specific length L (typically $L \gg P$)

$$w_r[n] = \begin{cases} w[rL + n] & ; 0 \leq n < L \\ 0 & ; \text{otherwise} \end{cases} \quad \text{duration interval } 0 \leq n < L$$

$$w[n] = \sum_{r=-\infty}^{\infty} w_r[n - rL], \forall n$$

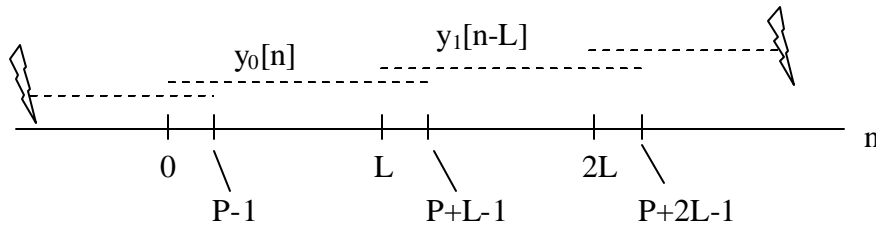
- $y_r[n] = h[n] * w_r[n]$ duration interval $0 \leq n < P + L - 1$
Find $y_r[n]$ by $(P+L-1)$ -point circular convolution

- $h[n] * w[n] = \sum_{r=-\infty}^{\infty} y_r[n - rL], \forall n$

Proof $h[n] * w[n] = h[n] * \sum_{r=-\infty}^{\infty} w_r[n - rL] = \sum_{r=-\infty}^{\infty} h[n] * w_r[n - rL]$

$$= \sum_{r=-\infty}^{\infty} y_r[n - rL]$$

Note that $h[n] * w_r[n - rL] = y_r[n - rL]$ from the time-invariance property of convolution.



- **Frequency-sampling approximation** of $h[n]$

To approximate $h[n]$:

- Sampling $\hat{H}(\mathbf{w})$ at $\mathbf{w} = k \frac{2\mathbf{p}}{N}$; $0 \leq k < N$ by

$$e^{jk \frac{2\mathbf{p}}{N} n} \xrightarrow{h[n]} \text{yield } \tilde{H}[k] = \hat{H}\left(\mathbf{w} = k \frac{2\mathbf{p}}{N}\right)$$

- Use $(\text{DFT})^{-1}$, $\tilde{h}[n] = \frac{1}{N} \sum_{k=0}^{N-1} \tilde{H}[k] \mathbf{y}_N^{nk} = \sum_{r=-\infty}^{\infty} h[n - rN]$; $0 \leq n < N$

$$\frac{(\hat{H}(\mathbf{w})|_{\mathbf{w}=\mathbf{w}_0}) e^{jn\mathbf{w}_0}}{e^{jn\mathbf{w}_0}} = (\hat{H}(\mathbf{w})|_{\mathbf{w}=\mathbf{w}_0})$$

$$\tilde{H}[k] = \hat{H}\left(\mathbf{w} = k \frac{2\mathbf{p}}{N}\right) \xrightarrow{\text{DFT}^{-1}} \tilde{h}[n] = \begin{cases} \sum_{r=-\infty}^{\infty} h[n - rN] & 0 \leq n < N \\ 0 & \text{otherwise} \end{cases}$$

- $y[n] = \sum_{r=-\infty}^{\infty} x[n - rN] \xleftrightarrow{\text{DFT}} \hat{Y}[k] = \hat{X}\left(\mathbf{w} = k \frac{2\mathbf{p}}{N}\right)$

Proof

$$\begin{aligned}
\tilde{h}[n] &= \frac{1}{N} \sum_{k=0}^{N-1} \tilde{H}[k] \mathbf{y}_N^{nk} = \frac{1}{N} \sum_{k=0}^{N-1} \hat{H}\left(k \frac{2\mathbf{p}}{N}\right) \mathbf{y}_N^{nk} \\
&= \frac{1}{N} \sum_{k=0}^{N-1} \left(\sum_{m=-\infty}^{\infty} h[m] e^{-jmk \frac{2\mathbf{p}}{N}} \right) e^{j \frac{2\mathbf{p}}{N} nk} \\
&= \sum_{m=-\infty}^{\infty} h[m] \left(\frac{1}{N} \sum_{k=0}^{N-1} e^{j \frac{2\mathbf{p}}{N} k(n-m)} \right) \\
\frac{1}{M} \sum_{\ell=0}^{M-1} e^{j2\mathbf{p} \ell \frac{n}{M}} &= \begin{cases} 1; & \text{if } \frac{n}{M} \in I \\ 0; & \text{if } \frac{n}{M} \notin I \end{cases} \Rightarrow \frac{1}{N} \sum_{k=0}^{N-1} e^{j2\mathbf{p}k \frac{n-m}{N}} = \begin{cases} 1; & \text{if } \frac{n-m}{N} \in I \\ 0; & \text{if } \frac{n-m}{N} \notin I \end{cases} \\
\tilde{h}[n] &= \sum_{\substack{m=-\infty \\ \frac{n-m}{N} \in I}}^{\infty} h[m]
\end{aligned}$$

So, the summation only include m of the form $m = n - rN$; $r \in I$

$$\text{Thus, } \tilde{h}[n] = \sum_{r=-\infty}^{\infty} h[n - rN]$$

$$\text{Proof 2: } \tilde{h}[n] = \sum_{r=-\infty}^{\infty} h[n - rN] \xrightarrow{DFT} \tilde{H}[k] = \hat{H}\left(\mathbf{w} = k \frac{2\mathbf{p}}{N}\right)$$

$$\text{First, note that } \hat{H}\left(k \frac{2\mathbf{p}}{N}\right) = \sum_{n=-\infty}^{\infty} h[n] e^{-jnk \frac{2\mathbf{p}}{N}} = \sum_{n=-\infty}^{\infty} h[n] \mathbf{y}_N^{-nk}.$$

$$\begin{aligned}
\tilde{H}[k] &= \sum_{n=0}^{N-1} \tilde{h}[n] \mathbf{y}_N^{-nk} = \sum_{n=0}^{N-1} \left(\sum_{r=-\infty}^{\infty} h[n + rN] \right) \mathbf{y}_N^{-nk} \\
&= \sum_{r=-\infty}^{\infty} \sum_{n=0}^{N-1} (h[n + rN] \mathbf{y}_N^{-nk})
\end{aligned}$$

Substitute $\ell = n + rN \Rightarrow n = \ell - rN$

$$\tilde{H}[k] = \sum_{r=-\infty}^{\infty} \sum_{\ell=rN}^{(r+1)N-1} (h[\ell] \mathbf{y}_N^{-(\ell-rN)k}) = \sum_{r=-\infty}^{\infty} \sum_{\ell=rN}^{(r+1)N-1} (h[\ell] \mathbf{y}_N^{-\ell k})$$

$$\text{Note that } \sum_{r=-\infty}^{\infty} \sum_{\ell=rN}^{(r+1)N-1} x(\ell) = \sum_{r=-\infty}^{\infty} \sum_{\ell=rN}^{rN+(N-1)} x(\ell) = \sum_{\ell=-\infty}^{\infty} x[\ell].$$

$$\text{Thus, } \tilde{H}[k] = \sum_{\ell=-\infty}^{\infty} h[\ell] \mathbf{y}_N^{-\ell k} = \hat{H}\left(k \frac{2\mathbf{p}}{N}\right).$$

- Time-aliasing

- If $h[n]$ has duration interval contained in $0 \leq n < N$,
 $\tilde{h}[n] = h[n]$; $0 \leq n < N$

- Example: if $h[n]$ is a $(2N)$ -point signal

$$\begin{aligned}
\hat{H}\left(k \frac{2p}{N}\right) &= \sum_{n=-\infty}^{\infty} h[n] e^{-jnk \frac{2p}{N}} = \sum_{n=-\infty}^{\infty} h[n] \mathbf{y}_N^{-nk} = \sum_{n=0}^{2N-1} h[n] \mathbf{y}_N^{-nk} \\
&= \sum_{n=0}^{N-1} h[n] \mathbf{y}_N^{-nk} + \sum_{n=N}^{2N-1} h[n] \mathbf{y}_N^{-nk} \\
&= \sum_{n=0}^{N-1} h[n] \mathbf{y}_N^{-nk} + \sum_{\ell=0}^{N-1} h[\ell + N] \mathbf{y}_N^{-(\ell+N)k} \\
&= \sum_{n=0}^{N-1} h[n] \mathbf{y}_N^{-nk} + \sum_{\ell=0}^{N-1} h[\ell + N] \mathbf{y}_N^{-\ell k} \\
&= \sum_{n=0}^{N-1} (h[n] + h[n + N]) \mathbf{y}_N^{-nk}
\end{aligned}$$

$$\tilde{h}[n] = h[n] + h[n + N]; 0 \leq n < N$$

So, $\tilde{h}[n] = h[n]; 0 \leq n < N$ iff $h[n] = 0$ for $n \geq N$ (no folding)

- **Windowing a signal** to get an approximation of $\hat{X}(\mathbf{w})$

Given $x[n] -\infty < n < \infty$

Look at $y[n] = x[n]p_L[n] = \begin{cases} x[n] & ; -L < n < L \\ 0 & ; |n| \geq L \end{cases}$ (length = $2L-1 = N$)

$$\text{Find } \hat{Y}(\mathbf{w}) = \frac{1}{2p} \int_{-p}^p \frac{\sin\left(\left(L - \frac{1}{2}\right)\mathbf{m}\right)}{\sin\left(\frac{\mathbf{m}}{2}\right)} \hat{X}(\mathbf{w} - \mathbf{m}) d\mathbf{m}$$

$$\bullet \quad p_L[n] = \begin{cases} 1 & ; -L < n < L \\ 0 & ; |n| \geq L \end{cases} \xleftrightarrow{\text{DFT}} \frac{\sin\left(\left(L - \frac{1}{2}\right)\mathbf{w}\right)}{\sin\left(\frac{\mathbf{w}}{2}\right)}$$

- Zeros @ $\frac{kp}{L - \frac{1}{2}}$
- = $2L-1$ @ $\mathbf{w} \rightarrow 0$

$$\bullet \quad \text{Need width of } \hat{X}(\mathbf{w}) \text{'s feature} > \frac{2p}{L - \frac{1}{2}} = \frac{4p}{N}$$

- If $L \rightarrow \infty$, $\lim_{L \rightarrow \infty} p_L[n] = 1 \xleftrightarrow{\text{DFT}} 2p\delta(\mathbf{w})$, and $\hat{Y}(\mathbf{w}) \rightarrow \hat{X}(\mathbf{w})$
- Lots of "sharp activity" in $\hat{X}(\mathbf{w}) \Rightarrow$ need bigger N
- Given $x[n]; 0 \leq n < N$

Can use M-point DFT's to find $\hat{X}[k]$ only for $k = \frac{N}{M}\ell = r\ell$; $0 \leq \ell < M$; $r = \frac{N}{M} \in I$

by let $y[n] = \sum_{p=0}^{r-1} x[n + (p-1)M]$, then $\hat{X}[k] = \hat{Y}[\ell]$

Proof

$$\begin{aligned}\hat{X}[k] &= \sum_{n=0}^{N-1} x[n] \mathbf{y}_N^{-nk} = \sum_{n=0}^{N-1} x[n] e^{-j\frac{2\mathbf{p}}{N}\left(\frac{N}{M}\ell\right)n} = \sum_{n=0}^{N-1} x[n] \mathbf{y}_M^{-\ell n} \\ &= \sum_{p=0}^{r-1} \sum_{n=pM}^{(p+1)M-1} x[n] \mathbf{y}_M^{-\ell n} = \sum_{p=0}^{r-1} \sum_{n=0}^{M-1} x[n + (p-1)M] \mathbf{y}_M^{-\ell(n + (p-1)M)} \\ &= \sum_{n=0}^{M-1} \sum_{p=0}^{r-1} x[n + (p-1)M] \mathbf{y}_M^{-\ell n}\end{aligned}$$

- Let $x[n]$ be a signal of duration N .
If already have tool for M-point DFT's.

Can find $\hat{X}(\mathbf{w})$ for $\omega_k = \omega_0 + k \frac{2\mathbf{p}}{M}$, by constructing $y[n]$ so $\hat{Y}[k] = \hat{X}(\mathbf{w}_k)$.

To do this,

- Let $q[n] = e^{-j\mathbf{w}_0 n} x[n]$, then $\hat{Q}(\mathbf{w}) = \hat{X}(\mathbf{w} + \mathbf{w}_0)$ and

$$\hat{Q}\left(k \frac{2\mathbf{p}}{M}\right) = \hat{X}\left(k \frac{2\mathbf{p}}{M} + \mathbf{w}_0\right) = \hat{X}(\mathbf{w}_k)$$

- Want $\hat{Y}[k] = \hat{Q}\left(k \frac{2\mathbf{p}}{M}\right)$,

From $y[n] = \sum_{r=-\infty}^{\infty} x[n - rN] \xrightarrow{DFT} \hat{Y}[k] = \hat{X}\left(\mathbf{w} = k \frac{2\mathbf{p}}{N}\right)$, so need

$$y[n] = \begin{cases} \sum_{r=-\infty}^{\infty} q[n + rM] & 0 \leq n < M \\ 0 & \text{otherwise} \end{cases}$$

- Note that not all r 's are used. only ones that satisfy

$$0 < n + rM < N \Rightarrow -\frac{n}{M} \leq r < \frac{N-n}{M}, \text{ but } 0 \leq n < M. \text{ So, } -1 < r < \frac{N}{M}$$

$$\Rightarrow 0 \leq r < \frac{N}{M}$$