Digital Signal Processing

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Z-Transform

Fourier Transform of a discrete time signal:

$$X(e^{jw}) = \sum_{k=-\infty}^{\infty} x(k)e^{-jwk}$$

Given a sequence x(n), its z transform is defined:

$$X(z) = \sum_{n=-\infty}^{\infty} x(n)z^{-n}$$

Where **z** is a **complex variable** $z=e^{j\omega}$

- The z transform does not converge for all sequences or for all values of z
- The set of values of z for which the z transform converges is called region of convergence
- The properties of the sequence x(n) determines the region of convergence of X(z)

Z TRANSFORM AND DFT Z-Transform

Finite-Length Sequences : FIR filters $X(z) = \sum_{n=0}^{\infty} x(n)z^{-n}$

$$X(z) = \sum_{n=n_1}^{n_2} x(n) z^{-n}$$

Convergence requires :

$$|x(n)| < \infty$$
 $n_1 \le n \le n_2$

$$n_1 \le n \le n_2$$

z may take all values **except** : $z = \infty$ if $n_1 < 0$

$$z=\infty$$
 if $n_1<0$

and
$$z = 0$$
 if $n_2 > 0$

Region of convergence : $0 < |z| < \infty$

$$0 < |z| < \infty$$

Compute X(z):

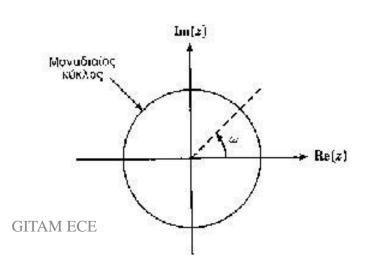
$$z = 1 (\omega = 0)$$

$$z = j (\omega = \pi/2)$$

$$z = -1 (\omega = \pi)$$

Unit circle inside

Region of convergence



Z TRANSFORM AND DFT Z-Transform

In many cases X(z) is a rational function:

Ratio of polynomials

Values of z for which X(z)=0 Zeros of X(z)

Values of **z** for which **X(z)=infinity Poles** of **X(z)**

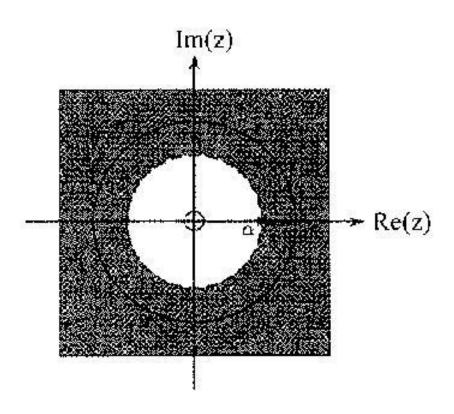
- No poles of X(z) can occur within the region of convergence (is bounded by poles)
- Graphically display z transform by pole-zero plot

Example:

Compute the **z** transform of the sequence $x(n)=a^nu(n)$

$$X(z) = \sum_{n=-\infty}^{\infty} a^n u(n) z^{-n} = \sum_{n=0}^{\infty} (az^{-1})^n =$$
 $= \frac{1}{1 - az^{\frac{1}{1-n}}} = \frac{z}{z-a}$

Z TRANSFORM AND DFT Z-Transform



- If |a|<1 the unit circle is included in the region of convergence, X(z) converges
- For causal systems X(z) converges everywhere outside a circle passing through the pole farthest from the origin of the plane.

Z-Transform

Ακολουθία	Μετασχηματισμός Ζ
$\delta(n)$	1
$\alpha^n u(n)$	$\frac{1}{1-\alpha z^{-1}}$
$-\alpha^n u(-n-1)$	$\frac{1}{1-\alpha z^{-1}}$
$n\alpha^n u(n)$	$\frac{\alpha z^{-1}}{(1-\alpha z^{-1})^2}$
$-n\alpha^n u(-n-1)$	$\frac{\alpha z^{-1}}{(1-\alpha z^{-1})^2}$
$\cos(n\omega_0)u(n)$	$\frac{1-(\cos\omega_0)z^{-1}}{1-2(\cos\omega_0)z^{-1}+z^{-2}}$
$\sin(n\omega_0)u(n)$	$(\sin \omega_0)z^{-1}$
	$1-2(\cos\omega_0)z^{-1}+z^{-2}$

Properties of the Z-Transform

1) Linearity:
$$ax_1(n) + bx_2(n) \Rightarrow aX_1(z) + bX_2(z)$$

2) Shifting:
$$x(n-m) \Rightarrow z^{-m}X(z)$$

3) Time scaling by a Complex Exponential Sequence :
$$a^n x(n) \Rightarrow X(a^{-1}z)$$

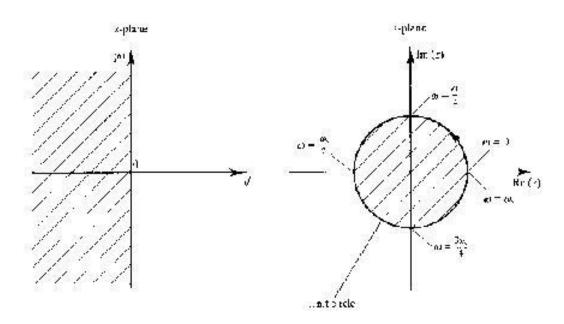
4) Convolution:
$$y(n) = x(n) * h(n) \Rightarrow Y(z) = X(z)H(z)$$

5) Differentiation:
$$nx(n) \Rightarrow -z \frac{dX(z)}{dz}$$

Relationship between Z-Transform and Laplace

- If $z=e^{sT}$, $s=d+j\omega$ $z=e^{(d+j\omega)T}=e^{dT}e^{j\omega T}$
- Then,

$$\mid z \mid = e^{dT}$$
 $\angle z = \omega T = 2\pi f/F_s = 2\pi \omega/\omega_s$



- Stability: Poles should be inside the unit circle
- **Stability** criterion: Finding the poles of the system
- FIR digital filters always stable: Poles in origin

Geometric Evaluation of Fourier Transform

- X(z) has M zeros at z=z₁,z₂,...,z_M
- X(z) has N poles at z=p₁,p₂,...,p_N
- We can write X(z) in factored form:

$$X(z) = A rac{\prod\limits_{i=1}^{M} (1-z_i z^{-1})}{\prod\limits_{i=1}^{N} (1-p_i z^{-1})}$$

Multiplying factors X(z) can be written as a rational fraction:

$$X(z) = rac{\sum\limits_{i=0}^{M} a_i z^{-i}}{1 + \sum\limits_{i=1}^{N} b_i z^{-i}}$$

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This form is often used for general filter design

Geometric Evaluation of Fourier Transform

The Fourier transform or system function:

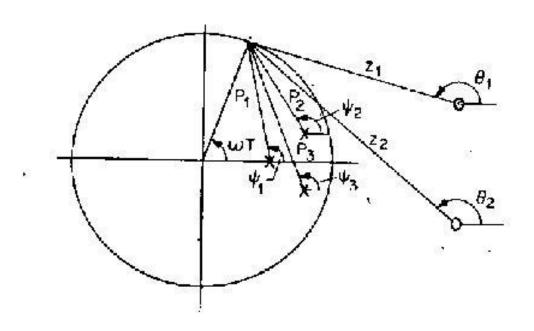
Evaluating X(z) on the unit circle, z=e^{jω}

$$X(e^{j\omega}) = A \frac{\prod\limits_{i=1}^{M} (1 - z_i e^{-j\omega})}{\prod\limits_{i=1}^{N} (1 - p_i e^{-j\omega})}$$

$$\left|egin{array}{c} X(e^{j\omega})
ight| = \left|A
ight| \left| egin{array}{c} 1 - z_i e^{-j\omega}
ight| & \prod\limits_{i=1}^M \left|e^{j\omega} - z_i
ight| \ X(e^{j\omega}) \left| = \left|A
ight| \left| egin{array}{c} rac{i=1}{N}
ight| 1 - p_i e^{-j\omega}
ight| & \prod\limits_{i=1}^M \left|e^{j\omega} - p_i
ight| \ \end{array}
ight|$$

$$\angle X(e^{j\omega}) = \angle A + \sum_{i=1}^{M} \angle (1 - z_i e^{-j\omega}) - \sum_{i=1}^{N} \angle (1 - p_i e^{-j\omega})$$

Z TRANSFORM AND DFT Geometric Evaluation of Fourier Transform



- From the point $z=e^{j\omega}$ draw vectors to zeros and poles
- Magnitudes of vectors determine magnitude at ω
- Angles determine phase

Example :
$$|X(e^{j\omega})| = \frac{Z_1Z_2}{P_1P_2P_3}$$

$$\angle X(e^{j\omega}) = heta_1 + heta_2 - (\psi_1$$
mec $\psi_2 + \psi_3)$

Inverse Z-Transform

From the inverse z transform we get x(n)

- Power series (long division)
- Partial fraction expansion
- Residue Theorem

Power series (long division)

X(z) can be written as rational fraction:

$$X(z) = \frac{a_0 + a_1 z^{-1} + a_2 z^{-2} + \ldots + a_N z^{-N}}{b_0 + b_1 z^{-1} + b_2 z^{-2} + \ldots + b_M z^{-M}}$$

It can be extended into an infinite series in z-1 by long division:

$$X(z) = x(0) + x(1)z^{-1} + x(2)z^{-2} + x(3)z^{-3} + \dots$$

Inverse Z-Transform

Example:

Find the first 4 values of the sequence f(k)

$$F(z) = \frac{2z^{-1}}{2z^{-2} - 3z^{-1} + 1} = \frac{2z}{z^2 - 3z + 2}$$

$$\frac{2z}{z^2 - 3z + 2} = 2z^{-1} + 6z^{-2} + 14z^{-3} + \cdots$$

$$f(k)=\{0,2,6,14....\}$$

The long division approach can be reformulated so x(n) can be obtained recursively:

ong division approach can be reformulated so
$$\mathbf{x}(\mathbf{n})$$
 can be $x(n) = \frac{\left[a_n - \sum\limits_{i=1}^n x(n-i)b_i\right]}{b_0}$ $n=1,2,...$ $x(0) = \frac{a_0}{b_0}$ GITAM ECE

Inverse Z-Transform

Partial fraction expansion:

$$X(z) = \frac{a_0 + a_1 z^{-1} + a_2 z^{-2} + \ldots + a_N z^{-N}}{b_0 + b_1 z^{-1} + b_2 z^{-2} + \ldots + b_M z^{-M}}$$

If poles of X(z) first order (distinct) and N=M,

$$X(z) = B_0 + \frac{C_1}{1 - p_1 z^{-1}} + \frac{C_2}{1 - p_2 z^{-1}} + \dots + \frac{C_M}{1 - p_M z^{-1}}$$

$$= B_0 + \frac{C_1 z}{z - p_1} + \frac{C_2 z}{z - p_2} + \dots + \frac{C_M z}{z - p_M}$$

- p(k): distinct poles, C_k partial fraction coef.
- $B_0=a_0/b_0$
- If N<M then B₀= 0
- If N>M then by long division make N<=M
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Inverse Z-Transform

• The coefficient C_k can be derived as:

$$C_k = \frac{X(z)}{z}(z - p_k) \bigg|_{z=p_k}$$

If X(z) contains multiple poles extra terms are required - X(z) contains mth-order poles:

$$\sum_{i=1}^m \frac{D_i}{(z-p_k)^i}$$

$$D_{i} = \frac{1}{(m-i)!} \frac{d^{m-i}}{dz^{m-i}} \Big[(z - p_{k})^{m} X(z) \Big] \bigg|_{z=p_{k}}$$

Inverse Z-Transform

Example:

Find the inverse z-transform:

$$X(z) = \frac{2z^{-1}}{2z^{-2} - 3z^{-1} + 1} = \frac{2z}{z^2 - 3z + 2} = \frac{2z}{(z - 2)(z - 1)}$$

$$\frac{X(z)}{z} = \frac{2}{(z - 2)(z - 1)} = \frac{C_1}{(z - 1)} + \frac{C_2}{(z - 2)}$$

$$C_1 = \frac{X(z)}{z}(z - 1) \Big|_{z=1} = \frac{2}{(1 - 2)} = -2$$

$$C_2 = \frac{X(z)}{z}(z - 2) \Big|_{z=2} = \frac{2}{(2 - 1)} = 2$$

$$Z^{-1} = \left[\frac{z}{(z - a)}\right] = a^n, \quad n \ge 0$$

$$Z^{-1} = \left[\frac{z}{(z - 1)}\right] = u(n), \quad n \ge 0$$
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$$x(n) = 2 \cdot 2^n - 2, \quad n \ge 0$$

Inverse Z-Transform

Residue Theorem

IZT obtained by evaluating the contour integral:

$$x(n) = \frac{1}{2\pi j} \oint_C z^{n-1} X(z) dz$$

• Where **C** is the path of integration enclosing all the poles of X(z).

Cauchy's residue theorem:

- Sum of the residues of z ⁿ⁻¹X(z) at all the poles inside C
- Every residue C_k, is associated with a pole at p_k

$$Res\Big[z^{n-1}X(z),p_k\Big] = \frac{1}{(m-1)!} \frac{d^{m-1}}{dz^{m-1}} \Big[(z-p_k)z^{n-1}X(z)\Big] \Big|_{z=p_k}$$

- m is the order of the pole at z=p_k
- For a **first-order** pole:

$$Resigg[z^{n-1}X(z),p_kigg]=(z-p_k)z^{n-1}X(z)igg|_{z=p_k}$$

Inverse Z-Transform

Example:

Find the **inverse z** transform :
$$X(z) = \frac{z^2}{(z-0.5)(z-1)^2}$$

$$z^{n-1}X(z) = \frac{z^{n+1}}{(z-0.5)(z-1)^2}$$

Single pole @ z=0.5, second-order pole @ z=1

$$Res \left[z^{n-1} X(z), 0.5 \right] = \frac{(z - 0.5) z^{n+1}}{(z - 0.5) (z - 1)^2} = \frac{z^{n+1}}{(z - 1)^2} \Big|_{z = 0.5}$$
$$= \frac{(0.5)^{n+1}}{(-0.5)^2} = 2(0.5)^n$$

$$egin{array}{lll} Res \Big[z^{n-1} X(z), 1 \Big] &=& rac{d}{dz} \Big[rac{(z-1)^2 z^{n+1}}{(z-0.5)(z-1)^2} \Big] \ &=& rac{(z-0.5)(n+1) z^n - z^{n+1}}{(z-0.5)^2} \, \Big|_{z=1} \ &=& rac{(0.5)(n+1) - 1}{(0.5)^2} = 2(n-1) \end{array}$$

Inverse Z-Transform

Combining the results we have:

$$x(n)=2[(n-1)+(0.5)^n]$$

No need to use inverse tables!!!

Comparison of the inverse z-transform

Power series:

Does not lead to a closed form solution, it is **simple**, easy computer implementation

- Partial fraction, residue:
 - Closed form solution,
 - Need to factorize polynomial (find poles of X(z))
 - May involve high order differentiation (multiple poles)
- Partial fraction: Useful in generating the coefficients of parallel structures for digital filters.
- * Residue method: widely used in the analysis of quantization errors in discrete-time systems.

Solving Difference Equations Using Z-Transform

The difference equation of interest (IIR filters) is:

$$y(n) = \sum_{i=0}^{N} b_i x(n-i) - \sum_{i=0}^{M} a_i y(n-i) \qquad n \ge 0$$

The z-transform is:

$$Y(z) = \sum_{i=0}^N b_i z^{-i} X(z) - \sum_{i=0}^M a_i z^{-i} Y(z)$$

Transfer function is:

$$H(z) = rac{Y(z)}{X(z)} = rac{\sum\limits_{i=0}^{N} b_i z^{-i}}{1 + \sum\limits_{i=0}^{M} a_i z^{-i}}$$

If coefficients $\mathbf{a_i} = \mathbf{0}$ (FIR filter): $H(z) = \frac{Y(z)}{X(z)} = \sum_{i=0}^{N} b_i z^{-i}$

Solving Difference Equations Using Z-Transform

Example:

Find the output of the following filter: y(n) = x(n) + a y(n-1)

Initial condition: y(-1) = 0

Input: $\mathbf{x}(\mathbf{n}) = \mathbf{e}^{\mathrm{j}\omega\mathbf{n}} \mathbf{u}(\mathbf{n})$

Using z transform:

$$Y(z) = X(z) + a z^{-1} Y(z)$$

$$x(n) = e^{jwn}, \qquad X(z) = \frac{1}{1 - e^{jw}z^{-1}}$$

$$H(z) = \frac{Y(z)}{X(z)} = \frac{1}{1 - az^{-1}}$$

$$Y(z) = H(z)X(z) = \frac{1}{(1 - az^{-1})(1 - e^{jw}z^{-1})}$$

Using partial fraction expansion:

$$Y(z) = \frac{a/(a-e^{jw})}{(1-az^{-1})} + \frac{-e^{jw}/(a-e^{jw})}{(1-e^{jw}z^{-1})}$$

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$$y$$
(n) $= \Big[rac{a^n}{a-e^{jw}} - rac{e^{jwn}}{a-e^{jw}}\Big]u(n)$

Discrete Fourier Transform (DFT)

- Techniques for representing sequences:
 - ❖ Fourier Transform
 - **❖ Z-transform**
 - Convolution summation
- Three good reasons to study DFT
 - It can be efficiently computed
 - Large number of applications
 - Filter design
 - Fast convolution for FIR filtering
 - Approximation of other transforms
 - Can be finitely parametrized
- When a sequence is **periodic** or **of finite duration**, the sequence can be represented in a **discrete-Fourier series**
- Periodic sequence x(n), period N,

$$x(n) = \sum_{k=-\infty}^{\infty} X(k) e^{j(2\pi/N)k\pi}$$

Discrete Fourier Transform (DFT)

- Remember: e^{jω} periodic with frequency 2π
- $2\pi k n / N = 2\pi n \rightarrow k = N$
- N distinct exponentials

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

,1/N just a scale factor

• The **DFT** is defined as:

$$X(k) = \sum_{n=0}^{N-1} x(n)e^{-j2\pi kn/N} = X(e^{jw}) \bigg|_{w=2\pi k/N}$$

• **DFT coefficients** correspond to **N** samples of **X(z)**:

$$X(k) = X(z)$$
 $\bigg|_{z=e^{j2\pi k/N}}$

Z TRANSFORM AND DFT Properties of the DFT

Linearity

If x(n) and y(n) are sequences (N samples) then:

$$a x(n) + b y(n) \leftarrow \rightarrow a X(k) + b Y(k)$$

Remember: x(n) and y(n) must be N samples,

otherwise zerofill

Symmetry

If **x(n)** is a **real** sequence of **N samples** then:

Re[X(k)] = Re [X(N-k)]
Im[X(k)] = -Im[X(N-k)]

$$|X(K)| = |X(N-k)|$$

Phase X(k) = - Phase X(N-k)

If x(n) is real and symmetric x(n) = x(N-n) then:

X(K) is purely real

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Properties of the DFT

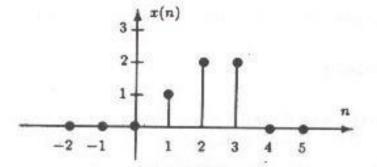
Shifting Property

If x(n) is periodic then $x(n) \leftarrow X(k)$,

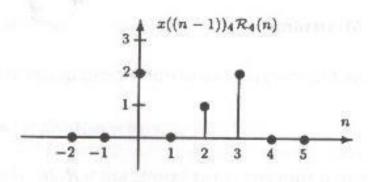
 $x(n-n_0) \leftarrow X(k) e^{-j(2\pi/N)} n_0^{k}$

If x(n) is not periodic then time-shift is created by rotating x(n) circularly by n_0

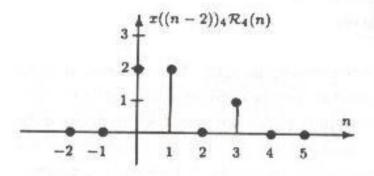
samples.



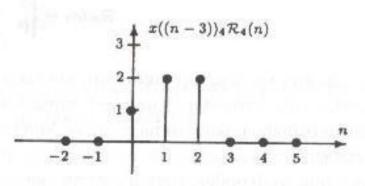
(α) Ένα σήμα διακριτού χρόνου μήκους N=4.



(β) Κυκλική μετατόπιση κατά ένα.



(γ) Κυκλική μετατόπιση κατά δύο.



(δ) Κυκλική μετατόπιση κατά τρία.

Convolution of Sequences

If x(n), h(n) are periodic sequences period N, DFTs:

$$X(k) = \sum_{n=0}^{N-1} x(n)e^{-j(2\pi/N)nk}$$

$$H(k) = \sum_{n=0}^{N-1} h(n)e^{-j(2\pi/N)nk}$$

$$H(k) = \sum_{n=0}^{N-1} h(n)e^{-j(2\pi/N)nk}$$

y(n): circular convolution of x(n), h(n)

Y(k) N-point DFT of y(n)

$$y(n) = \sum_{l=0}^{N-1} x(l)h(n-l) = x(n) \otimes h(n)$$

$$Y(k) = X(k) \cdot H(k)$$

Linear convolution has infinite sum.

$$y(n) = \sum_{l=-\infty}^{\infty} x(l) h(n-l)$$

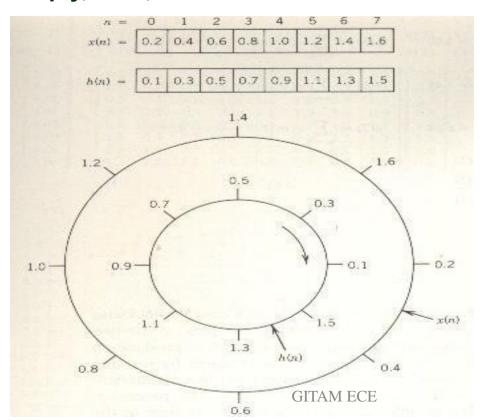
Convolution of Sequences

- Imagine one sequence around a circle N points.
- Second sequence around a circle N points but timed reversed

Convolution: multiply values of 2 circles, shift

multiply, shift, N times

Example:



Convolution of Sequences

$$n = -1 - 2 - 3 - 4 - 5 - 6 - 7 0 1 2 3 4 5 6 7 ... \\ x(n) 0.4 0.6 0.8 1.0 1.2 1.4 1.6 0.2 0.4 0.6 0.8 1.0 1.2 1.4 1.6 ... \\ h(-m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(2 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(3 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(4 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(6 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(7 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(8 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(9 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.5 0.3 0.1 \\ h(0 - m) 1.5 1.3 1.1 0.9 0.7 0.$$

Sectioned Convolution

- Fast Convolution: Using DFT for 2 finite sequences
 - Evaluated Rapidly, efficiently with FFT
 - $N_1+N_2 > 30 \rightarrow$ Fast Convolution more efficient
- Direct Convolution: direct evaluation
- $L > N_1 + N_2$ Add zeros to achieve L power of 2

Sectioned Convolution

- $N_1 \gg N_2$,what to do?
- L > N₁+N₂, inefficient and impractical. Why?
- Long sequence must be available before convolution
 - Practical waveforms: Speech, Radar not available
 - ❖ No processing before entire sequence Long delays
- Solution: Sectioned Convolution
- Overlap Add
- Overlap Save

Sectioned Convolution

Overlap - Add

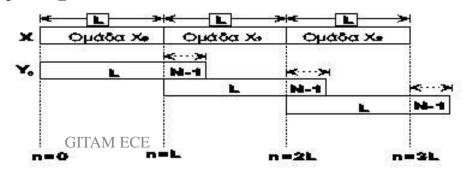
- Long sequence x(n) infinite duration
- Short sequence h(n) N₂ duration
- x(n) is sectioned N₃ or L or M

$$x(n) = \sum_{k=0}^{\infty} x_k(n)$$

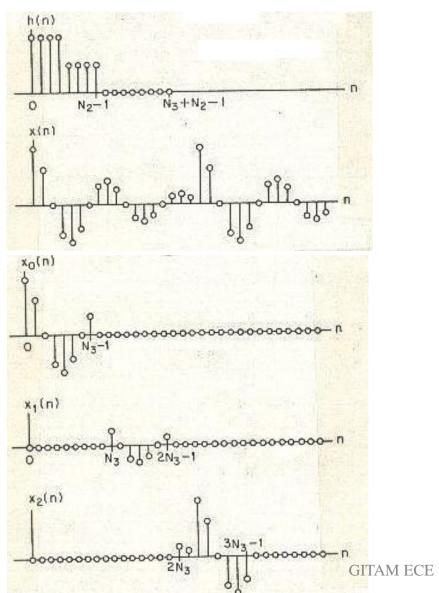
$$x_k(n) = egin{cases} x(n) & kN_3 \leq n \leq (k+1)N_3 - 1, \ 0 & ext{allows} \end{cases}$$

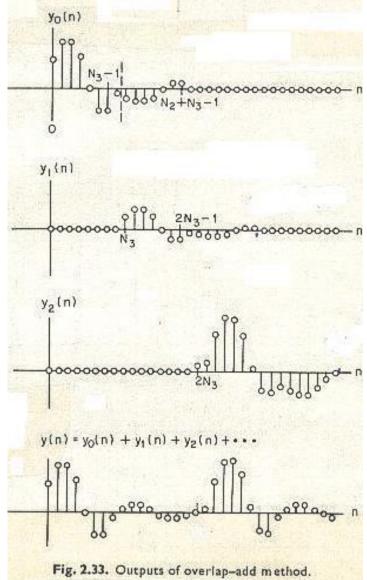
$$y(n) = \sum_{m=0}^{n} h(m) \sum_{k=0}^{\infty} x_k(n-m) = \sum_{k=0}^{\infty} y_k(n)$$

Duration of each convolution N₃ + N₂ - 1 (overlap)



Sectioned Convolution





Thank You