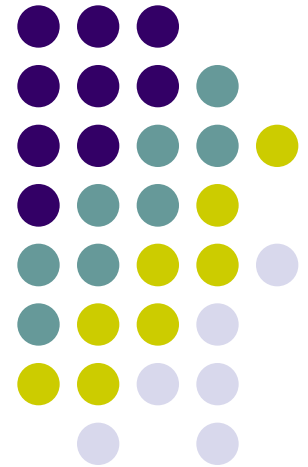


Discrete-Time signals and systems

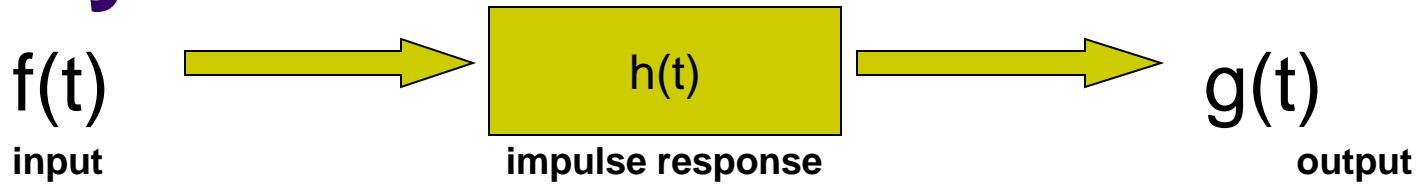


Introduction

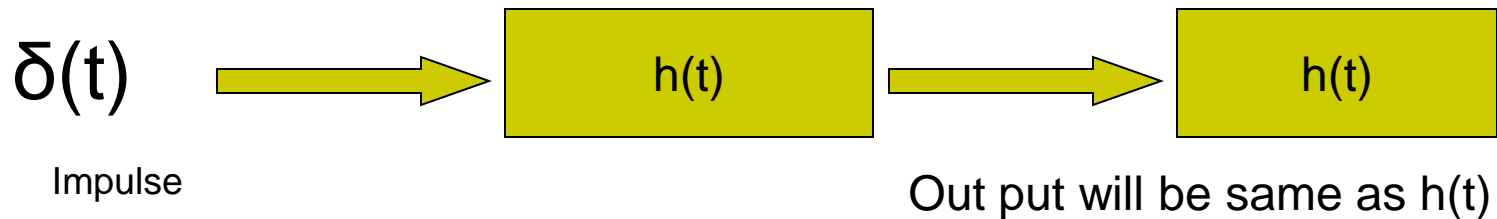


- **Signal:** A signal can be defined as a function that conveys information, generally about the state or behavior of a physical system.
- **Continuous-time signal:** Continuous-time signals are defined along a continuum of times and thus are represented by a continuous independent variable. Continuous-time signals are often referred to as analog signals.
- **Discrete time signal:** Discrete-time signals are defined at discrete times and thus the independent variable has discrete values; i.e., discrete-time signals are represented as sequences of number.

Analog: Linear time invariant system



• How to retrieve $h(t)$:





How to define $\delta(t)$?

- $\delta(t) = \begin{cases} \alpha & t = 0 \\ 0 & \text{Others} \end{cases}$

And $\int_{-\varepsilon}^{+\varepsilon} \delta(t) dt = 1$

How to define System Output $g(t)$?



- $g(t) = \int_{-\infty}^{+\infty} h(\tau) \cdot f(t - \tau) d\tau$
- Or $g(t) = \int_{-\infty}^{+\infty} f(\tau) \cdot h(t - \tau) d\tau$
- And there is another a form like
 $g(t) = h(t) * f(t)$

Here $*$ is the commutative property of convolution



Frequency domain:

- In frequency domain system output

$$G(\omega) = \int_{-\infty}^{+\infty} g(t) e^{-j\omega t} dt$$

$$G(\omega) = H(\omega) \times F(\omega)$$

Definition: Important functions in Discrete Time Signal



- Unit impulse function:

$$\delta(n) = \begin{cases} 1 & \text{for, } n = 0 \\ 0 & \text{for, } n \neq 0 \end{cases}$$

Unit step function:

$$U(n) = \begin{cases} 1 & \text{for, } n \geq 0 \\ 0 & \text{for, } n < 0 \end{cases}$$

Here “n” is an integer

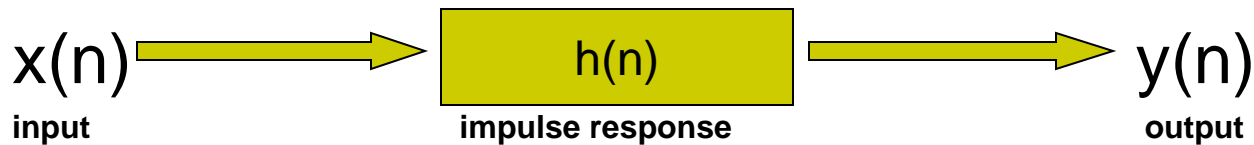
Relationship between $\delta(n)$ and $U(n)$



- $U(n) = \sum_{k=-\infty}^n \delta(k)$
- $U(n) = \sum_{k=0}^{\infty} \delta(n-k)$
- $\delta(n) = U(n) - U(n-1)$



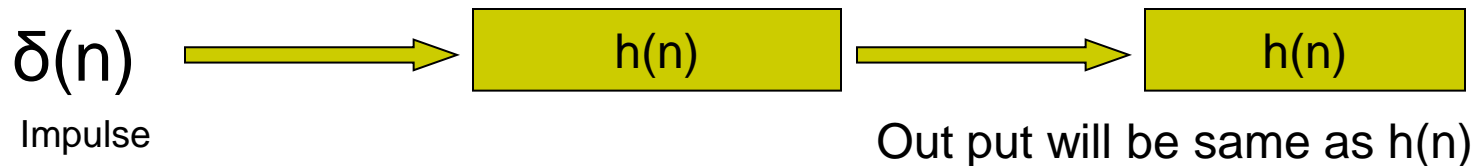
Discrete time invariant system



$$\text{Output } y(n) = \sum_{i=-\infty}^{i=+\infty} h(i)x(n-i)$$

$$\text{Or } y(n) = \sum_{i=-\infty}^{i=+\infty} x(i)h(n-i)$$

How to retrieve $h(n)$?

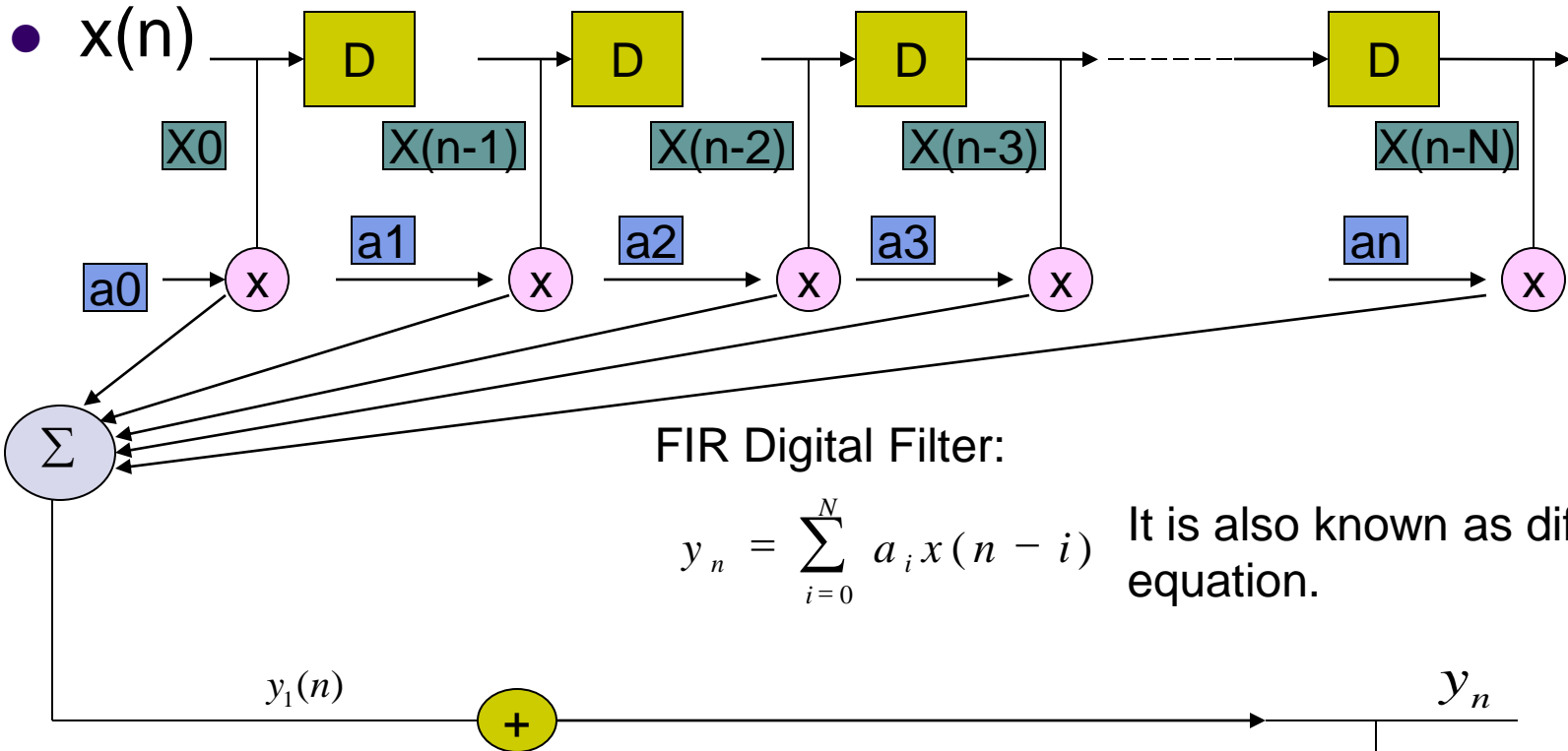




Definitions:

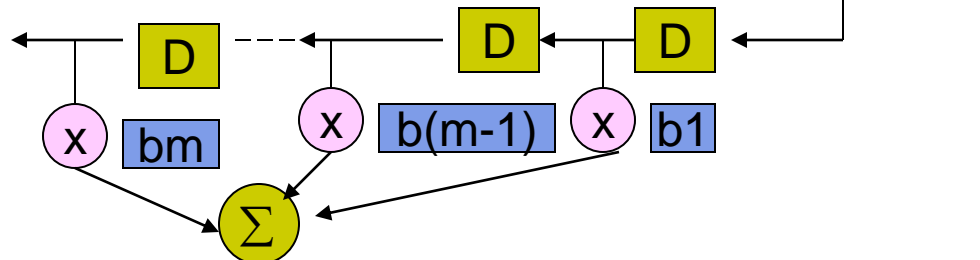
- **FIR (Finite Impulse Response):** A finite impulse response (FIR) filter is a type of a digital filter. The impulse response, the filter's response to a Kronecker delta input, is 'finite' because it settles to zero in a finite number of sample intervals.
- **IIR(Infinite Impulse Response):** They have an impulse response function which is non-zero over an infinite length of time.
- **Casual System:** If $h(n)=0$ for $n<0$; then this system is called casual system.

Follow graph(FIR digital filter,IIR digital filter):



IIR Digital Filter:

$$y_n = \sum_{i=0}^N a_i x(n - i) + \sum_{j=1}^m b_j y(n - j)$$





The System Output $h(n)$

$$h(n) = a_0 \delta(n) + a_1 \delta(n-1) + b_1 h(n-1)$$

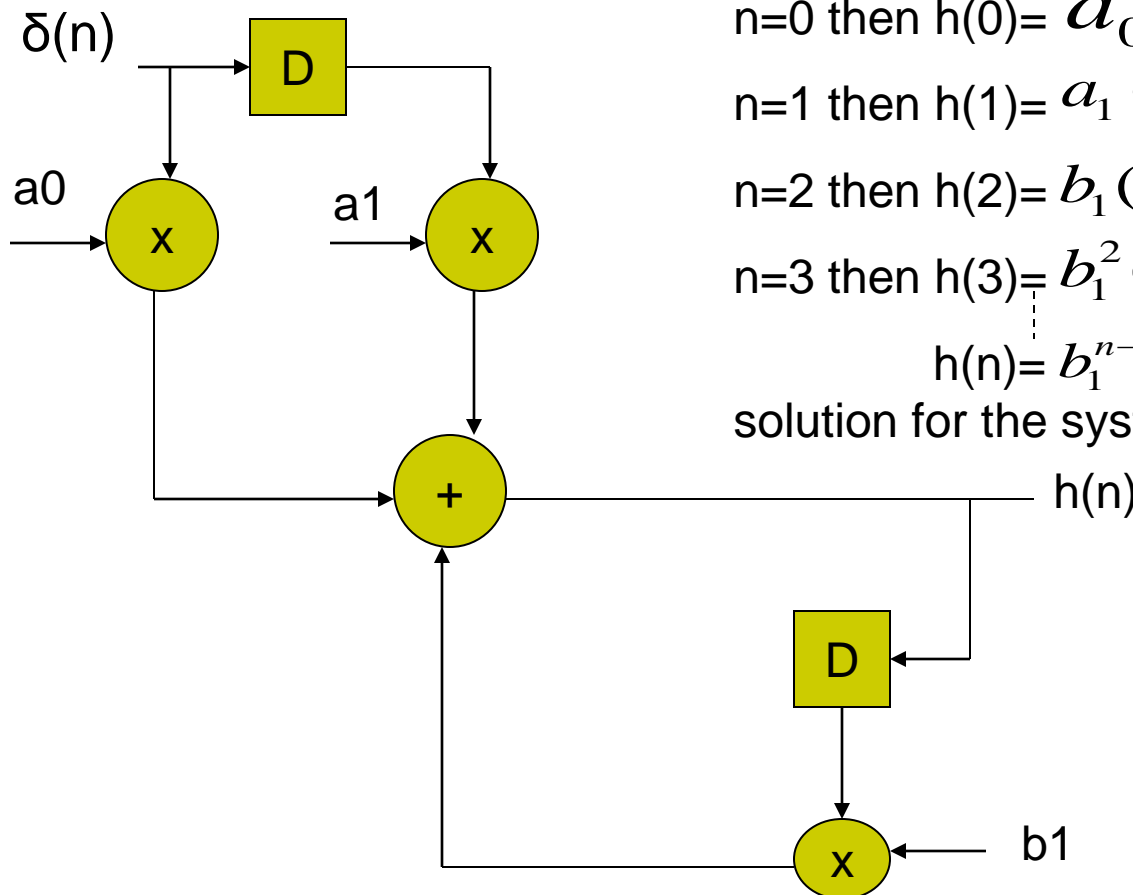
$$n=0 \text{ then } h(0) = a_0$$

$$n=1 \text{ then } h(1) = a_1 + b_1 a_0$$

$$n=2 \text{ then } h(2) = b_1 (a_1 + b_1 a_0)$$

$$n=3 \text{ then } h(3) = b_1^2 (a_1 + b_1 a_0)$$

$$h(n) = b_1^{n-1} (a_1 + b_1 a_0) \quad \text{it is general solution for the system output } h(n)$$



The energy of a sequence:

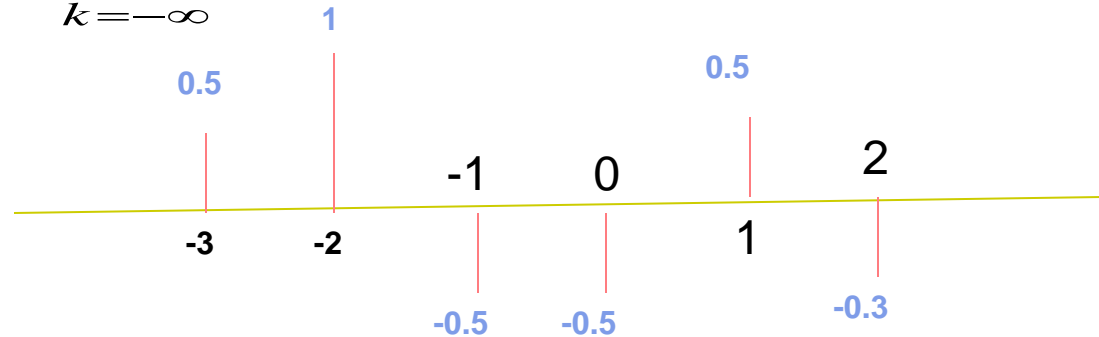


$$E = \sum_{n=-\infty}^{+\infty} |x(n)|^2$$

Any discrete sequence can be shown by $\delta(n)$:

General equation: $x(n) = \sum_{k=-\infty}^{\infty} x(k) \delta(n - k)$

Example:

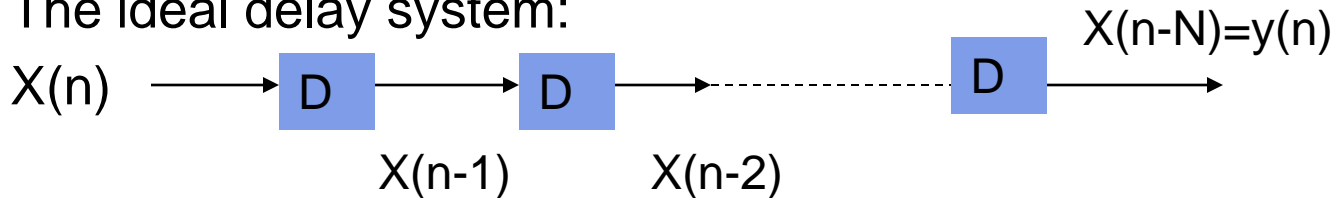


$$X(n) = 0.5\delta(n+3) + \delta(n+2) - 0.5\delta(n+1) - 0.5\delta(n) + 0.5\delta(n-1) - 0.3\delta(n-2)$$



Definition: Different System

The ideal delay system:



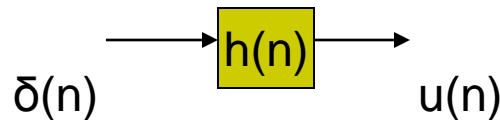
The ideal delay system output would be $y(n)=x(n-N)$

Moving Average: The moving average system output

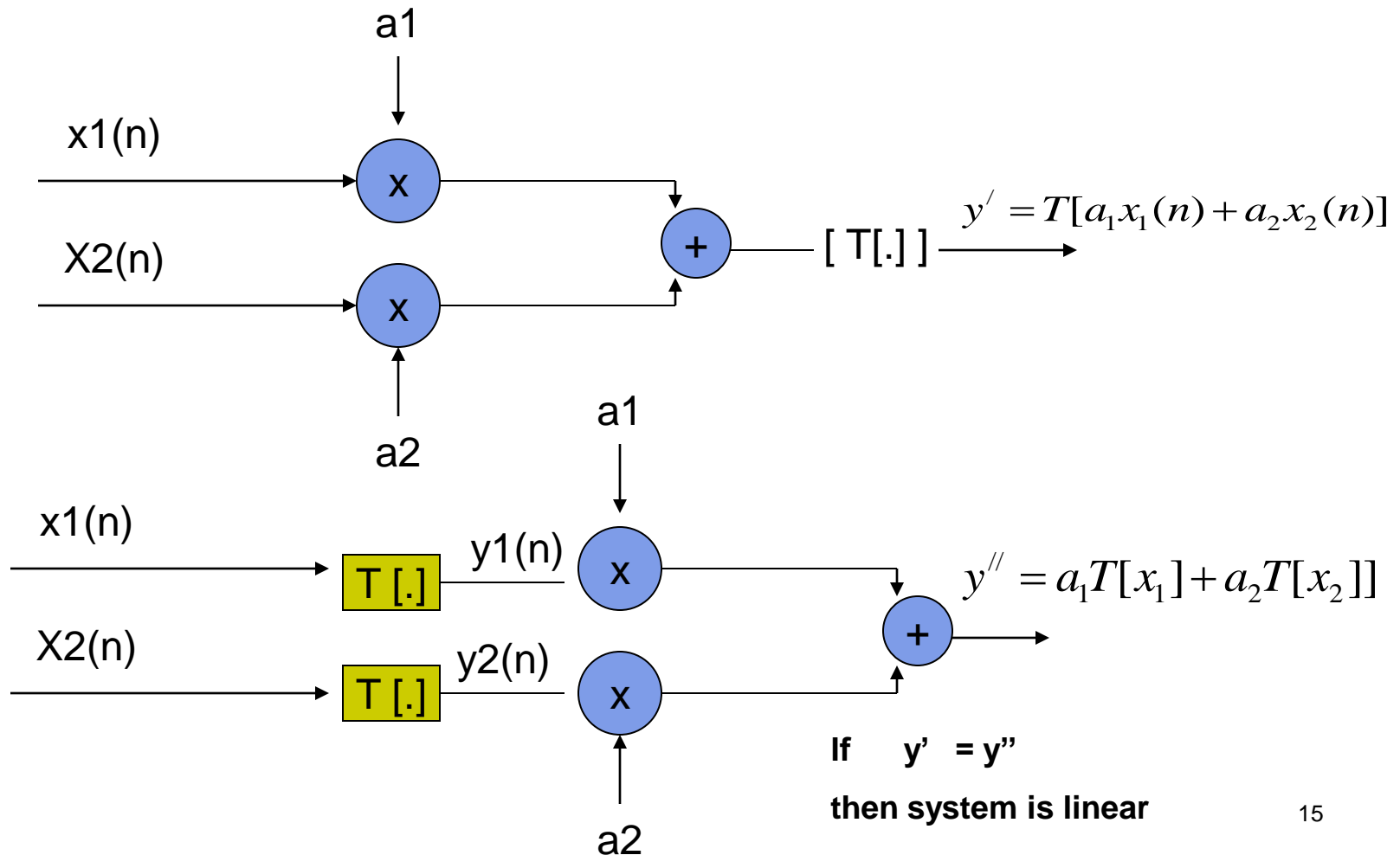
$$y(n) = \frac{1}{m_1 + m_2 + 1} \sum_{k=-m_1}^{m_2} x(n-k)$$

Accumulator (Acc) : $y(n) = \sum_{k=-\infty}^n x(n)$

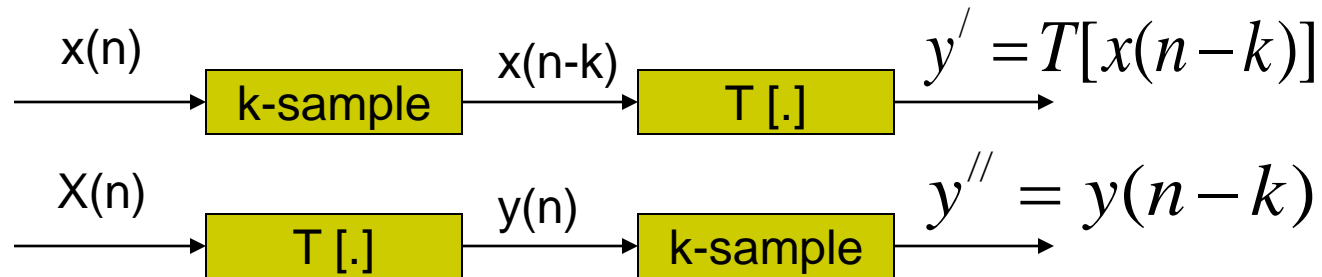
So the impulse response of the accumulator is same as the $u(n)$



Linear Shift Invariant System



Linear Shift Invariant System (Cont...)

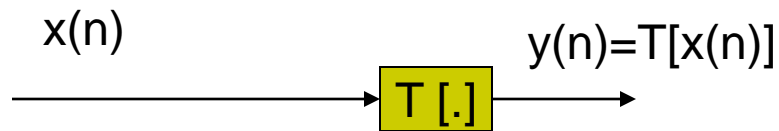


If $y'' = y(n-k) = T[x(n-k)]$ then the system is shift invariant



Discrete Convolution

- In LSI system



$$x(n) = \sum_{k=-\infty}^{+\infty} x_{(k)} \hat{\partial}_{(n-k)}$$

$$\text{So, } y(n) = T \left[\sum_{k=-\infty}^{+\infty} x_{(k)} \hat{\partial}_{(n-k)} \right]$$

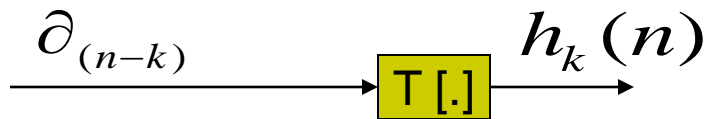
If the system is linear then

$$y(n) = \sum_{k=-\infty}^{+\infty} x_{(k)} T[\hat{\partial}_{(n-k)}]$$



Discrete Convolution (Cont...)

- If the system is LSI



So we can write

$$y(n) = \sum_{k=-\infty}^{+\infty} x_{(k)} h_{k(n)}$$

Again if the system is SI: $y_{(n-k)} = T[x_{(n-k)}]$ and $h_{k(n)} = h_{(n-k)}$

if LSI: $y(n) = \sum_{k=-\infty}^{+\infty} x_{(k)} h_{k(n-k)}$, it is known as discrete convolution

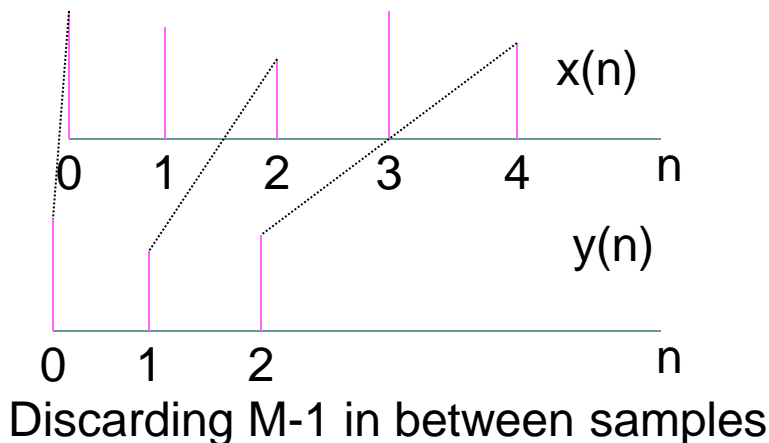
We can also write as $y(n) = \sum_{k=-\infty}^{+\infty} h_{(k)} x_{k(n-k)}$

Compressor: Down Sampling :Decimator



- The compressor output $y(n)=x(M.n)$ where M is a integer greater than 1

If $M=2$;



A compressor is not SI:

$$x_1(n)=x(n-n_0)$$

$$y_1(n)=x_1(mN)=x(Mn-n_0)$$

$$y(n-n_0)=x[M(n-n_0)]=x(Mn-Mn_0)$$

Here, $x(Mn-n_0) \neq x(Mn-Mn_0)$

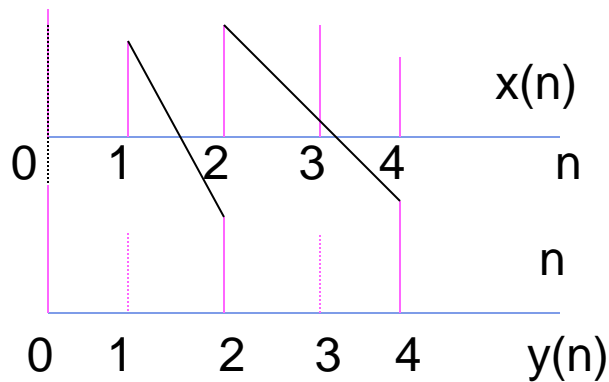
So, a compressor not SI



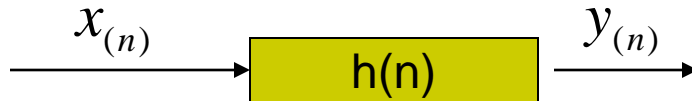
Expander : Up Sampling : Interpolation



- The expander output $y(n)=x(n/L)$; where $L>1$
- If $L=2$;



The system output $y(n)$ in different cases



- $h(n) = \begin{cases} a^n & \text{for } n \geq 0 \\ 0 & \text{for } n < 0 \end{cases}$

$x(n) = U(n) - U(n-N)$

$y(n) = ?$

We know that system output for LSI: $y(n) = \sum_{k=-\infty}^{+\infty} x_{(k)} h_{(n-k)}$

1. For $n < 0$ then $y(n) = 0$

2. For $0 \leq n < N$ then $y(n) = \sum_{k=0}^n a^{n-k} = a^n \sum_{k=0}^n a^{-k} = a^n \frac{1 - a^{-(n+1)}}{1 - a^{-1}}$

3. For $n \geq N-1$ then $y(n) = \sum_{k=0}^{N-1} a^{n-k} = a^n \frac{1 - a^{-N}}{1 - a^{-1}}$



Definition:

- **Stability of a system:** A stable system produces finite output when the input is finite

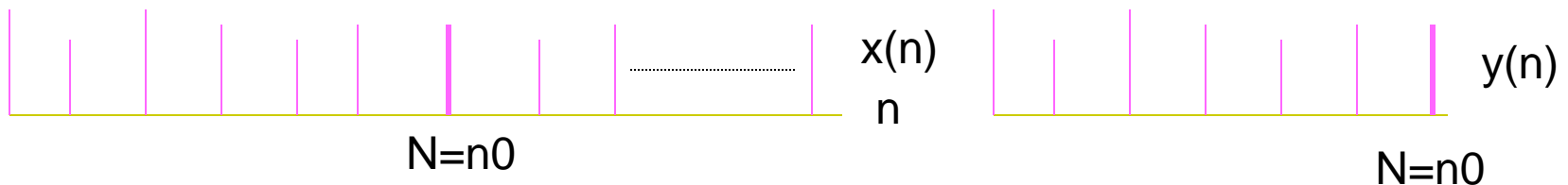
$$|y(n)| = \left| \sum_{k=-\infty}^{+\infty} h_k x_{(n-k)} \right| < \infty$$

For LSI system:

$$\text{and } |x(n)| < M < \infty$$

$$\text{So, } |y(n)| < M \sum_{k=-\infty}^{+\infty} |h_{(k)}| < \infty \Rightarrow \sum_{k=-\infty}^{+\infty} |h_{(k)}| < \infty$$

Causality: A system is casual when for $N=n_0$; the output of the system depending on input $x(n)$ only for $n \leq n_0$



If the system is LSI and $h(n)=0$ for $n < 0$ then it is causal

Example: $s = \sum_{n=-\infty}^{+\infty} |h(n)| = \sum_{n=0}^{\infty} |a^n| < \infty$

1. $|a| < 1$; $S = 1/1 - |a|$; Stable

2. $a = 1$ and $a > 1$; $s = \infty$ then not stable

Impulse response for different delay



Ideal Delay:



Impulse response for ideal delay $h(n) = \delta(n-m)$

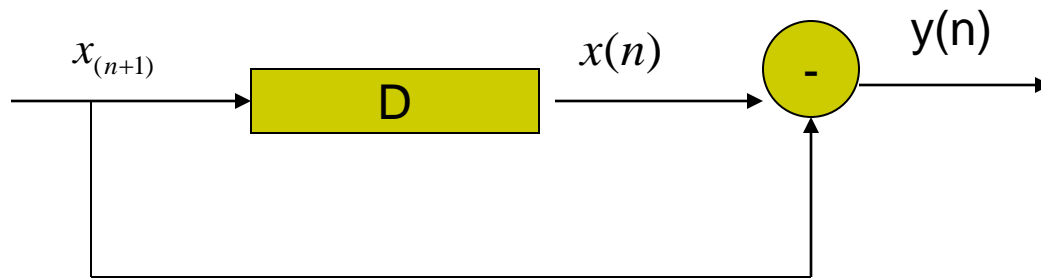
Moving average:

$$h(n) = \frac{1}{m_1 + m_2 + 1} \sum_{k=-m_1}^{m_2} \delta(n-k) = \begin{cases} \frac{1}{m_1 + m_2 + 1} & \text{for } n - m_1 \leq n \leq m_2 + n \\ 0 & \text{otherwise} \end{cases}$$

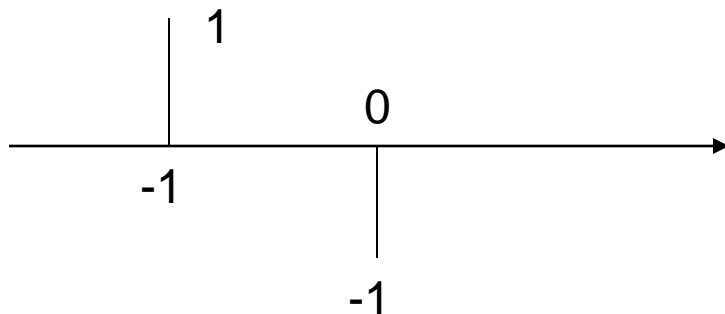
Accumulator: $h(n) = u(n) = \sum_{k=-\infty}^n \delta(k) = \begin{cases} 1 & n \geq 0 \\ 0 & n < 0 \end{cases}$



Forward difference:

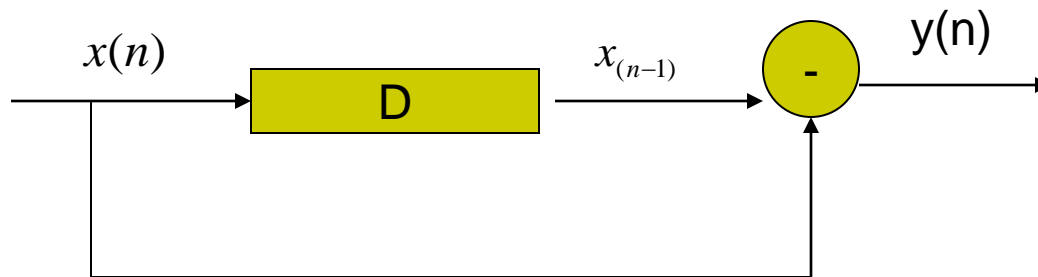


- Difference equation = $x(n+1) - x(n)$
- Impulse response = $\delta(n+1) - \delta(n)$

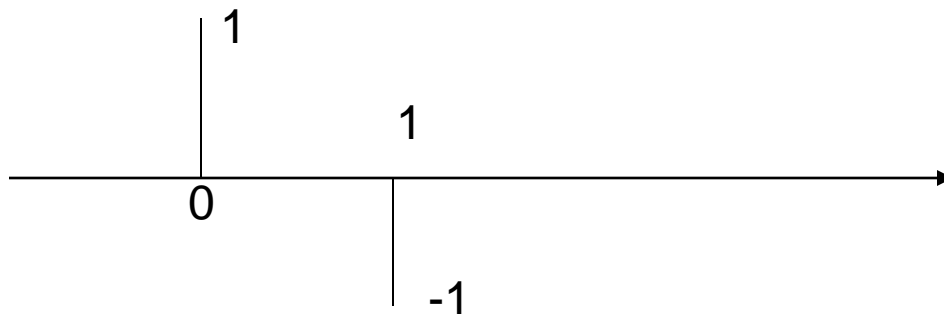




Backward Difference



- Difference equation = $y(n) = x(n) - x(n-1)$
- Impulse response $h(n) = \delta(n+1) - \delta(n)$





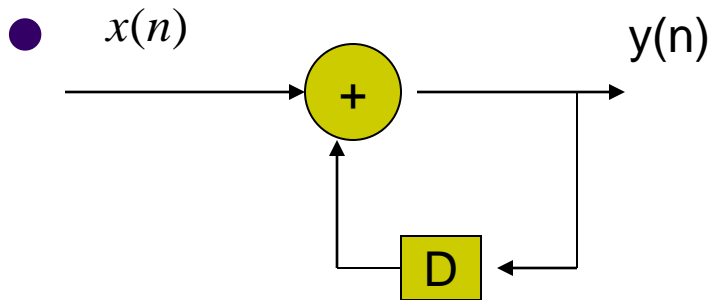
Stability:

- For a LSI $s = \sum_{n=-\infty}^{+\infty} |h(n)| < \infty$
- For Ideal Delay, Moving Average, Backward Difference and Forward Difference; if $s < \infty$ then it is stable

But for Accumulator

$s = \sum_{n=0}^{\infty} U(n)$ goes to ∞ then it is not stable

Accumulator

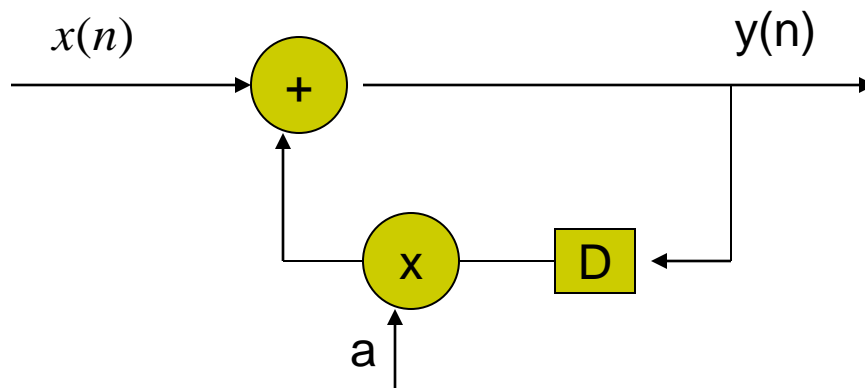


Difference equation

$$y(n) = x(n) + y(n-1)$$

$$h(n) = U(n)$$

- **This IIR digital filter**



Difference equation

$$y(n) = x(n) + ay(n-1)$$

Impulse response $h(n) = a^n u(n)$

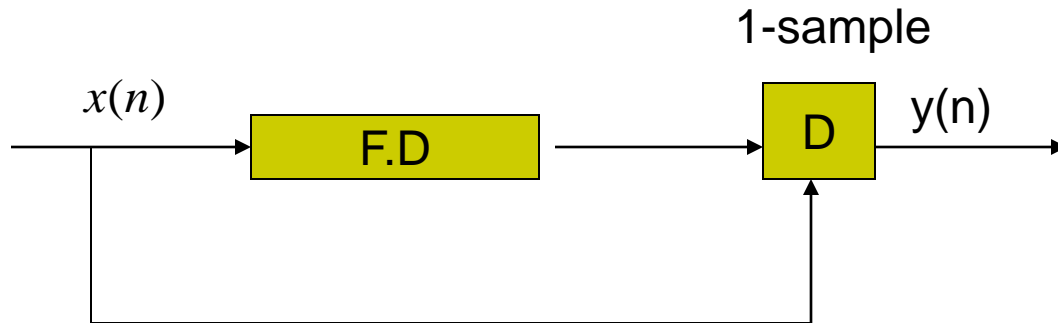
Stability checking $= \sum |a^n| = \frac{1}{1-|a|} < \infty$

Condition check: if $|a| < 1$

Result: Stable



Other properties of LSI

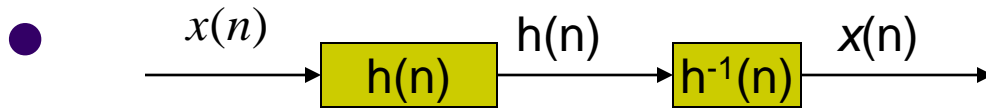


$$h(n)=[\delta(n+1)-\delta(n)]*\delta(n-1)$$

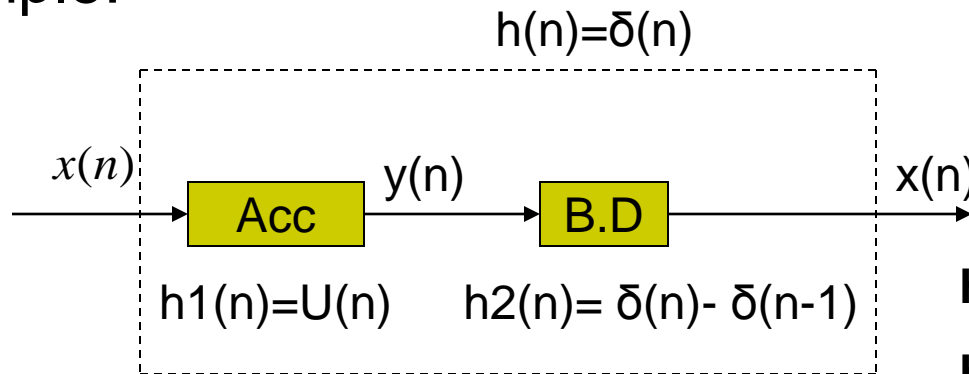
$h(n)=\delta(n)-\delta(n-1)$; In case of forward delay, if we add a 1-sample delay then it will convert to B.D



Inverse of a system

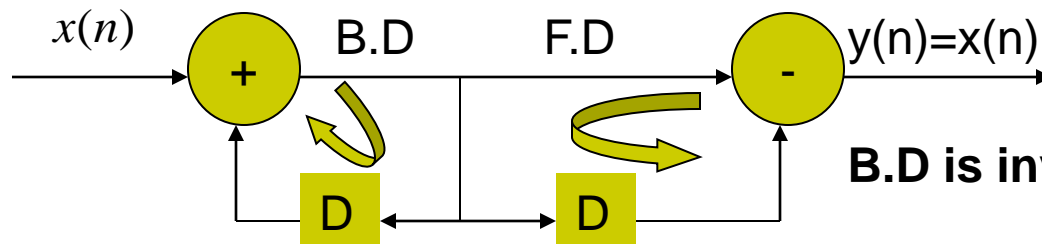


Example:



$$h(n) = h_1(n) * h_2(n)$$

$$h(n) = U(n) * [\delta(n) - \delta(n-1)] = U(n) - U(n-1) = \delta(n)$$



B.D is inverse system of ACC

Inverse of system: Engineering application



- 1. T.V.ghost canceling
- 2. Channel multi-path canceling
- 3. Equalization in communications channel

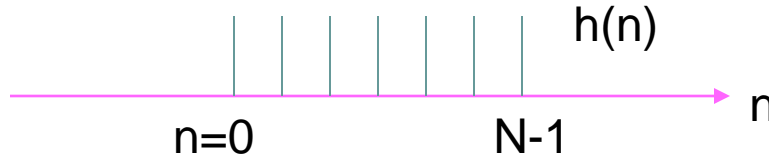


Frequency domain representation of discrete time signals and system

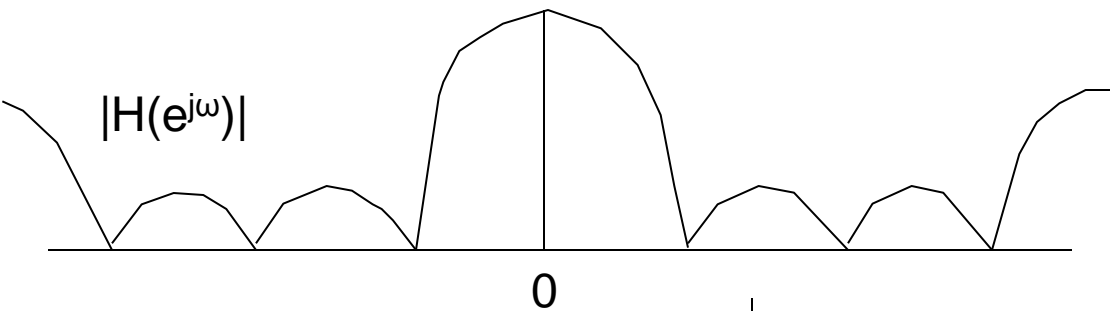
- Frequency response of the system:

$$H(e^{j\omega}) = \sum_{k=-\infty}^{+\infty} h(k)e^{-j\omega k}$$

Example:



$$h(n) = \begin{cases} 1 & 0 \leq n \leq N-1 \\ 0 & \text{elsewhere} \end{cases}$$

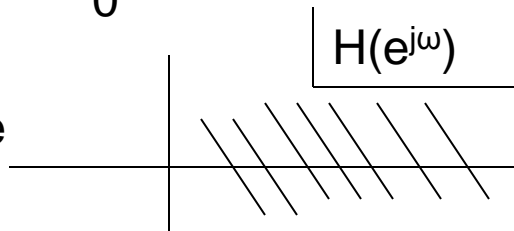


$$h(n) = U(n) - U(n-N)$$

$$H(e^{j\omega}) = \sum_{k=0}^{N-1} h(k)e^{-j\omega k} = \frac{1 - e^{-j\omega N}}{1 - e^{-j\omega}}$$

$$= \frac{\sin\left(\frac{\omega N}{2}\right)}{\sin\left(\frac{\omega}{2}\right)} e^{-j\frac{N-1}{2}\omega}$$

For linear phase

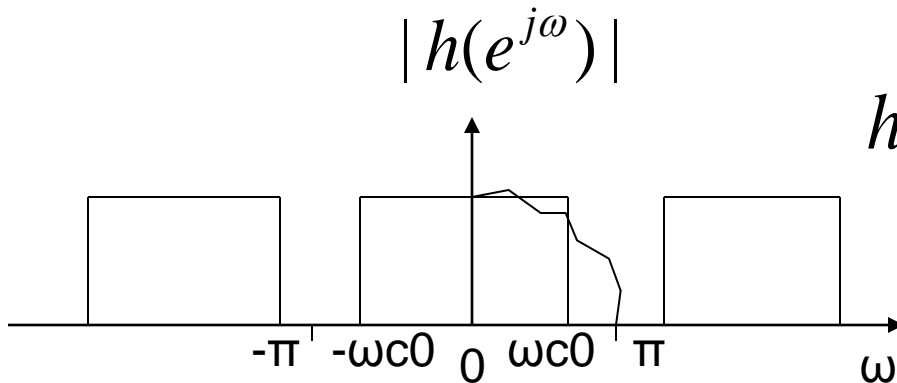




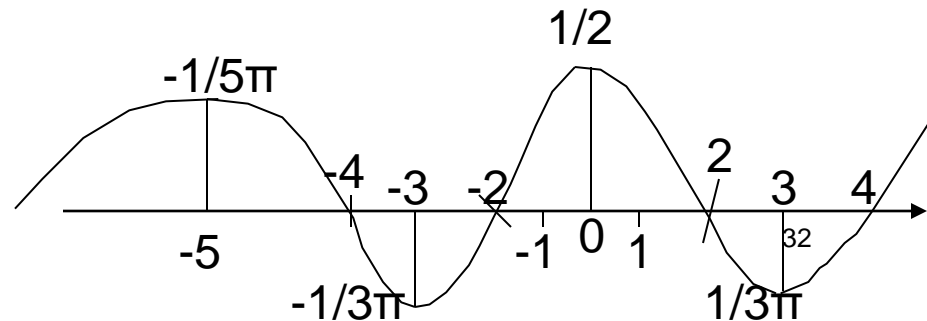
Inverse Fourier transform:

$$h(n) = \frac{1}{2\pi} \int_{-\pi}^{\pi} H(e^{j\omega}) e^{j\omega n} d\omega$$

Example: In the ideal low pass filter $H(e^{j\omega}) = \begin{cases} 1 & |\omega| \leq \omega_{c0} \\ 0 & \omega_{c0} < |\omega| \leq \pi \end{cases}$



$$h(n) = \frac{1}{2\pi} \int_{-\omega_{c0}}^{\omega_{c0}} e^{j\omega n} d\omega = \frac{\sin(\omega_{c0} n)}{\pi n}$$





Proof the inverse Fourier transform

$$x(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x_{(n)} e^{-j\omega n}$$

$$\therefore x_n = \frac{1}{2\pi} \int_{-\pi}^{\pi} x(e^{j\omega}) e^{j\omega n} d\omega$$

$$\hat{x} \text{ such that: } \hat{x}(n) = \frac{1}{2\pi} \int_{-\pi}^{\pi} \left(\sum_{m=-\infty}^{+\infty} x_{(m)} e^{-j\omega m} \right) e^{j\omega n} d\omega$$

interchanging the order of integral and summation

$$\hat{x}(n) = \sum_{m=-\infty}^{+\infty} x_{(m)} \cdot \left[\frac{1}{2\pi} \int_{-\pi}^{\pi} e^{j\omega(n-m)} d\omega \right]$$

Proof the inverse Fourier transform cont..



We have,

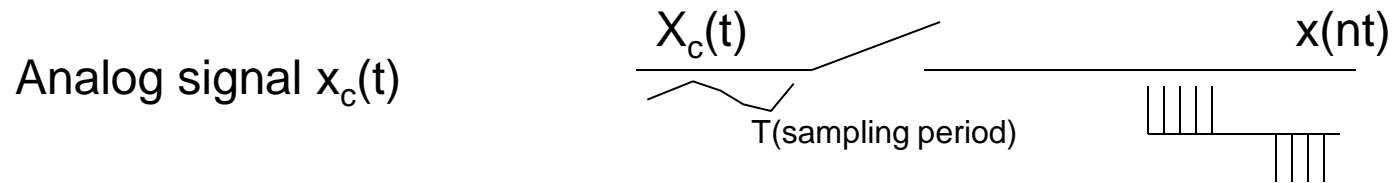
$$\frac{1}{2\pi} \int_{-\pi}^{\pi} e^{j\omega(n-m)} d\omega = \frac{\sin[\pi(n-m)]}{\pi(n-m)} = \begin{cases} 1 & \text{for } m=n \\ 0 & \text{for } m \neq n \end{cases}$$
$$= \delta(n)$$

$$\hat{x}(n) = \sum_{m=-\infty}^{+\infty} x(m) \delta(n-m)$$

$$\hat{x}(n) = x(n)$$



Sampling theorem:



$$\text{FT: } x_c(\Omega) = \int_{-\infty}^{+\infty} x_c(t) e^{-j\Omega t} dt \dots \dots \dots (1)$$

$$\text{IFT: } x_c(t) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} x_c(\Omega) e^{j\Omega t} d\Omega \dots \dots \dots (2)$$

Fourier transform for discrete time signal:

$$\text{Digital freq: } \omega = \Omega \cdot T \dots \dots \dots (3)$$



$$x(e^{j\omega}) = \sum_{n=-\infty}^{+\infty} x(nT)e^{-jn\omega} = \sum_{n=-\infty}^{+\infty} x(nT)e^{-jn\Omega T} \dots\dots\dots(4)$$

From eq.(2) for t=n.T

$$x(nT) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} x_c(\Omega)e^{j\omega nT} d\Omega \dots\dots\dots(5)$$

On the other hand we had before:

$$x(nT) = \frac{1}{2\pi} \int_{-\pi}^{+\pi} x(e^{j\omega})e^{j\omega n} d\omega \dots\dots\dots(6)$$



Putting (5) into (4)

$$x(e^{j\omega}) = \sum_{n=-\infty}^{+\infty} \frac{1}{2\pi} \int_{-\infty}^{+\infty} x_c(\Omega) e^{j\Omega nT} d\Omega e^{-jn\omega} \dots\dots\dots(7)$$

Interchanging Σ with integral

$$x(e^{j\omega}) = \frac{1}{2\pi} \int_{-\infty}^{\infty} x_c(\Omega) \left[\sum_{n=-\infty}^{\infty} e^{jn(\Omega T - \omega)} \right] d\Omega \dots\dots\dots(8)$$



But we have:

$$\sum_{n=-\infty}^{+\infty} e^{jn(\Omega T - \omega)} = \frac{2\pi}{T} \sum_{k=-\infty}^{+\infty} \delta\left(\Omega - \frac{\omega}{T} + \frac{2\pi k}{T}\right) \dots \dots \dots (9)$$

Then from (8) and (9) :

$$x(e^{j\omega}) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} x_c(\Omega) \cdot \left[\frac{2\pi}{T} \sum_{k=-\infty}^{+\infty} \delta\left(\Omega - \frac{\omega}{T} + \frac{2\pi k}{T}\right) \right] d\Omega$$

$$x(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{+\infty} \int_{-\infty}^{+\infty} x_c(\Omega) \delta\left(\Omega - \frac{\omega}{T} + \frac{2\pi k}{T}\right) d\Omega \dots \dots \dots (10)$$



Then

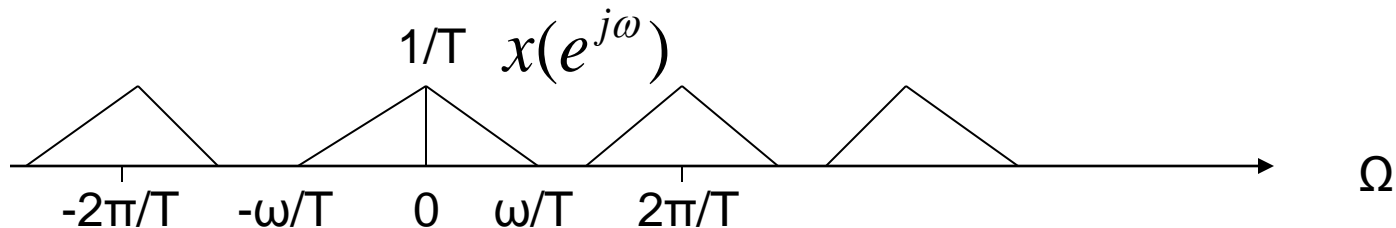
$$x(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{+\infty} x_c \left(\frac{\omega}{T} - \frac{2\pi k}{T} \right)$$

Since,

$$\Omega = \frac{\omega}{T}$$

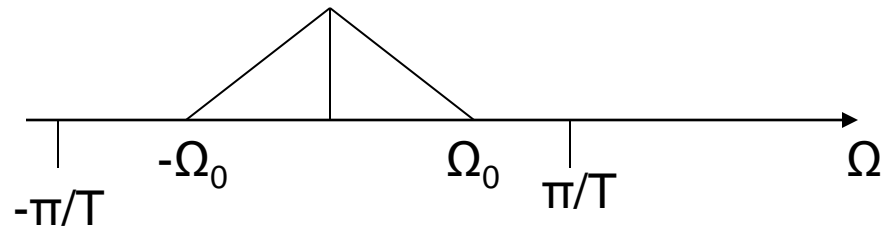
$$x(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{+\infty} x_c \left(\Omega - \frac{2\pi k}{T} \right) \dots \dots \dots (11)$$

Then $x(e^{j\omega})$ is a periodic function of $\Omega = \omega/T$ with period of $2\pi/T$





$x(\Omega)$: Analog spectrum



Nyquist rate for maximum frequency Ω_0 is sampling rate in order not to have aliasing effect

$$\text{if } \Omega_0 < \frac{\pi}{T} = \frac{\omega_s}{2} = \pi f_s$$

$$-\pi \leq \omega \leq \pi$$



In this case we can recover the analog signal from its sample as follows:

$$x_c(t) = \sum_{k=-\infty}^{+\infty} x_c(kt) \frac{\sin\left[\frac{\pi}{T}(t - kT)\right]}{\left[\frac{\pi}{T}(t - kT)\right]}$$

This formal is obtained as follows:

$$x_c(t) = \frac{1}{2\pi} \int_{-\pi/T}^{\pi/T} x_c(\Omega) e^{j\Omega t} d\Omega$$

$$-\frac{\pi}{T} \leq \Omega \leq \frac{\pi}{T}$$



$$x(e^{j\omega}) = x(e^{j\omega T}) = \frac{1}{T} x_c(\Omega)$$

$$x_c(t) = \frac{1}{2\pi} \int_{-\pi/T}^{+\pi/T} T \cdot x(e^{j\omega}) e^{j\Omega t} d\Omega$$

$$x(e^{j\omega}) = x(e^{j\Omega T}) = \sum_{k=-\infty}^{+\infty} x(kT) e^{-j\Omega T k}$$

$$x_c(t) = \frac{T}{2\pi} \int_{-\pi/T}^{+\pi/T} \left[\sum_{k=-\infty}^{+\infty} x(kT) e^{-j\Omega T k} \right] e^{j\Omega t} d\Omega$$



$$x_c(t) = \sum_{k=-\infty}^{+\infty} x(kT) \left[\frac{T}{2\pi} \int_{-\pi/T}^{\pi/T} e^{j\Omega(t-kT)} d\Omega \right]$$

$$x_c(t) = \sum_{k=-\infty}^{+\infty} x(kT) \frac{\sin \left[\left(\frac{\pi}{T} \right) (t - kT) \right]}{\left(\frac{\pi}{T} \right) (t - kT)}$$

To recover analog signal from its sample



Fourier transform properties

$$x(n) \xrightarrow{f} X(e^{j\omega})$$

1. Time shift: $x(n-M) \xrightarrow{f} e^{-j\omega M} X(e^{j\omega})$

2. Frequency shift: $e^{j\omega_0 n} x(n) \xleftarrow{f} X(e^{j(\omega-\omega_0)})$

3. Time reversal: $x(-n) \xrightarrow{f} X(e^{-j\omega})$

if $x(n)$ is a real sequence then:

$$x(-n) \xrightarrow{f} X^*(e^{j\omega})$$

Fourier transform properties contd..



4. Differentiations in frequency domain:

$$n \cdot x(n) \longrightarrow j \frac{dx(e^{j\omega})}{d\omega}$$

5. Convolution theorem:

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

$$Y(e^{j\omega}) = X(e^{j\omega}) \cdot H(e^{j\omega})$$

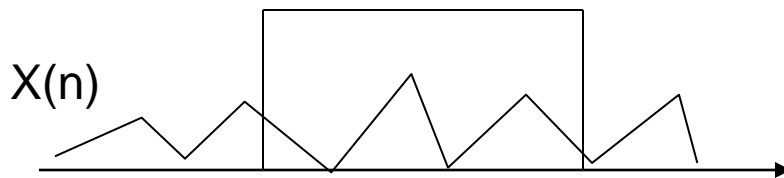
Fourier transform properties contd..



6. Parseval's Theorem (Energy)

$$E = \sum_{n=-\infty}^{+\infty} |x(n)|^2 = \frac{1}{2\pi} \int_{-\pi}^{\pi} |x(e^{j\omega})|^2 d\omega$$

7. The modulation or windowing theorem:



$$y(n) = w(n) \cdot x(n)$$

$$Y(e^{j\omega}) = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\theta}) \cdot w(e^{j(\omega-\theta)}) d\theta$$



Z-Transform

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

z: complex variable in z-plane

Similar to Laplace transform

Convergency of Z-transfer should be check

Example 1. $x(n) = a^n U(n)$

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

From the geometric series if we have $|az^{-1}| < 1 \Rightarrow |z| > |a|$
 $x(z) = 1/1 - az^{-1}$



Example

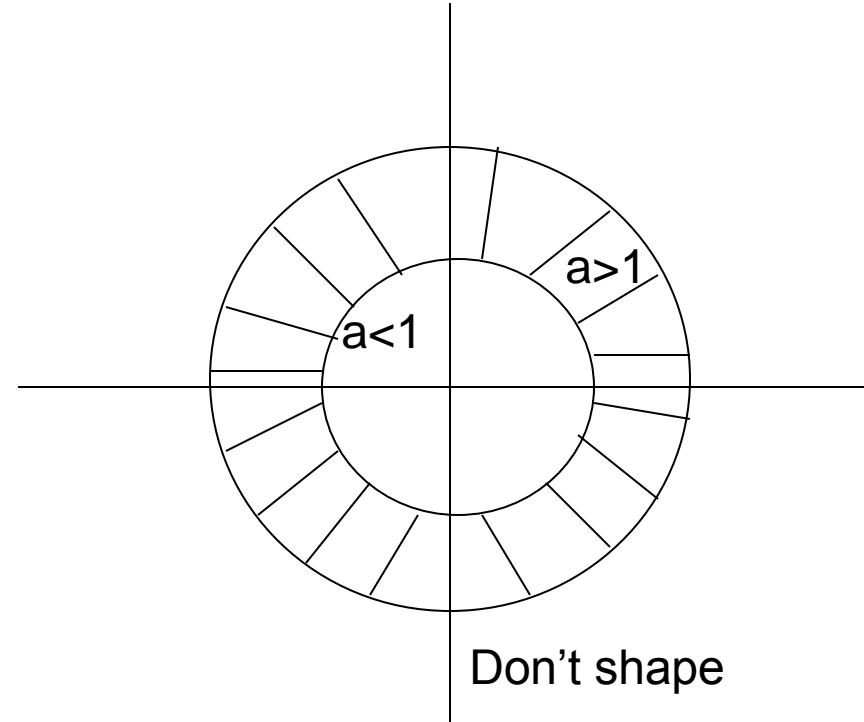
- $x(n)=a^{|n|}$
 $|a|<1$

$$\begin{aligned} X(z) &= \sum_{n=1}^{\infty} a^{-n} z^{-n} + \sum_{n=0}^{\infty} a^n z^{-n} \\ &= \sum_{n=1}^{\infty} a^n z^n + \sum_{n=0}^{\infty} a^n z^{-n} \\ &\quad \quad \quad 1 \qquad \qquad \quad 2 \end{aligned}$$

1- convergent for $|az|<1$

2-convergent for $|az^{-1}|<1$

where $|a|<|z|<1/|a|$





Example contd...

So,

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

$$Z(z) = \frac{1-a^z}{(1-az)(1-az^{-1})}$$



Example

- Example 2:

$$x(n) = \delta(n)$$

$$X(z) = 1$$

$$x(n) = \delta(n-m)$$

$$X(z) = z^{-m}$$

Example 3: $x(n) = a^n \sin(\omega_0 n) U(n)$

$$X(z) = \frac{a \sin(\omega_0) z^{-1}}{1 - 2a \cos \omega_0 z^{-1} + a^2 z^{-2}}$$



Example

- Example 4
 $x(n) = na^n U(n)$

$$X(z) = \sum_{n=0}^{\infty} na^n z^{-n}$$

With a little change in above summation

$$X(z) = z^{-1} \frac{d}{dz^{-1}} \left[\sum_{n=0}^{\infty} a^n z^{-n} \right] = z^{-1} \frac{d}{dz^{-1}} \left[\frac{1}{1 - az^{-1}} \right]$$

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



Properties of z-transform

- 1. Linearity: $a_1x_1(n)+a_2x_2(n) \xrightarrow{\text{Z-trans}} a_1X_1(z)+a_2X_2(z)$

- 2. Shift: $x(n \pm k) \longrightarrow Z^{\pm k} X(z)$

- 3 Convolution: $y(n)=\sum_{k=-\infty}^{+\infty} h(k)x(n-k)$

$$Y(Z)=h(Z).X(Z)$$

The relation between Z transform to Fourier transform



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

If we put: $z=e^{j\omega}$ then Z.T \longrightarrow F.T

For more general case $Z=re^{j\omega}$

So,
$$X(re^{j\omega}) = \sum_{n=-\infty}^{+\infty} [x(n)r^{-n}] e^{-j\omega n}$$

z-transform derived from Laplace transform



Consider a discrete-time signal $x(t)$ below sampled every T sec

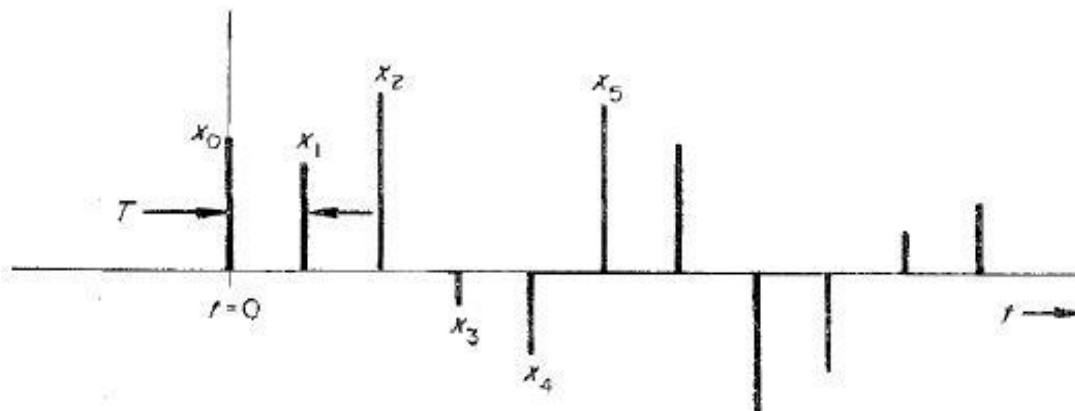
$$x(t) = x_0 \delta(t) + x_1 \delta(t - T) + x_2 \delta(t - 2T) + x_3 \delta(t - 3T) + \dots$$

The Laplace transform of $x(t)$ is therefore:

$$X(s) = x_0 + x_1 e^{-sT} + x_2 e^{-s2T} + x_3 e^{-s3T} + \dots$$

$$\text{Now define } z = e^{sT} = e^{(\sigma + j\omega)T} = e^{\sigma T} \cos \omega T + je^{\sigma T} \sin \omega T$$

$$X[z] = x_0 + x_1 z^{-1} + x_2 z^{-2} + x_3 z^{-3} + \dots$$





Range of convergency (ROC)

- $x(n)=u(n)$

in case $z=e^{j\omega}$ not convergent

in case $z=re^{j\omega}$ for $r>1$ then convergent because

$$\begin{aligned} & \sum_{n=-\infty}^{+\infty} |x(n)r^{-n}| < \infty \\ &= \sum_{n=-\infty}^{+\infty} |u(n)r^{-n}| < \infty \\ &= \sum_{n=0}^{\infty} |r^{-n}| < \infty \end{aligned}$$

ROC Contd....



- Step function $u(n)$ has z-transform for ROC: $|z|>1$
 - ∴ If ROC includes the unit circle then $z=e^{j\omega}$ and the sequence has Fourier Transform
 - ∴ There is possibility that two sequences are different but they may have a similar algebraic form of their z-transform, however their ROC's are different

Table of z-transform



Here:

- $u[n]=1$ for $n \geq 0$, $u[n]=0$ for $n < 0$
- $\delta[n] = 1$ for $n=0$, $\delta[n] = 0$ otherwise

	Signal, $x[n]$	Z-transform, $X(z)$	ROC
1	$\delta[n]$	1	all z
2	$\delta[n - n_0]$	z^{-n_0}	$z \neq 0$
3	$u[n]$	$\frac{1}{1 - z^{-1}}$	$ z > 1$
4	$-u[-n - 1]$	$\frac{1}{1 - z^{-1}}$	$ z < 1$
5	$nu[n]$	$\frac{z^{-1}}{(1 - z^{-1})^2}$	$ z > 1$
6	$-nu[-n - 1]$	$\frac{z^{-1}}{(1 - z^{-1})^2}$	$ z < 1$
7	$n^2u[n]$	$\frac{z^{-1}(1 + z^{-1})}{(1 - z^{-1})^3}$	$ z > 1$
8	$-n^2u[-n - 1]$	$\frac{z^{-1}(1 + z^{-1})}{(1 - z^{-1})^3}$	$ z < 1$
9	$n^3u[n]$	$\frac{z^{-1}(1 + 4z^{-1} + z^{-2})}{(1 - z^{-1})^4}$	$ z > 1$
10	$-n^3u[-n - 1]$	$\frac{z^{-1}(1 + 4z^{-1} + z^{-2})}{(1 - z^{-1})^4}$	$ z < 1$

Table of z-transform



11	$a^n u[n]$	$\frac{1}{1 - az^{-1}}$	$ z > a $
12	$-a^n u[-n - 1]$	$\frac{1}{1 - az^{-1}}$	$ z < a $
13	$na^n u[n]$	$\frac{az^{-1}}{(1 - az^{-1})^2}$	$ z > a $
14	$-na^n u[-n - 1]$	$\frac{az^{-1}}{(1 - az^{-1})^2}$	$ z < a $
15	$n^2 a^n u[n]$	$\frac{az^{-1}(1 + az^{-1})}{(1 - az^{-1})^3}$	$ z > a $
16	$-n^2 a^n u[-n - 1]$	$\frac{az^{-1}(1 + az^{-1})}{(1 - az^{-1})^3}$	$ z < a $
17	$\cos(\omega_0 n) u[n]$	$\frac{1 - z^{-1} \cos(\omega_0)}{1 - 2z^{-1} \cos(\omega_0) + z^{-2}}$	$ z > 1$
18	$\sin(\omega_0 n) u[n]$	$\frac{z^{-1} \sin(\omega_0)}{1 - 2z^{-1} \cos(\omega_0) + z^{-2}}$	$ z > 1$
19	$a^n \cos(\omega_0 n) u[n]$	$\frac{1 - az^{-1} \cos(\omega_0)}{1 - 2az^{-1} \cos(\omega_0) + a^2 z^{-2}}$	$ z > a $
20	$a^n \sin(\omega_0 n) u[n]$	$\frac{az^{-1} \sin(\omega_0)}{1 - 2az^{-1} \cos(\omega_0) + a^2 z^{-2}}$	$ z > a $

The properties of ROC



- ROC has a ring form or a disc form
- The Fourier transform of $x(n)$ has Fourier transform if and only if that its z-transform's ROC includes unit circle
- ROC cannot contain any pole
- If the sequence $x(n)$ has finite length then ROC contains all z-plane (excluding $z=0$ or $z=\infty$)
- If $x(n)$ is right-sided, then ROC is located outside of the largest pole.
- If $x(n)$ is left sided then ROC is located inside of the smallest pole.
- If the sequence $x(n)$ is both-sided then the ROC has ring shape which is limited to inside and outside poles and there is no pole in ROC.
- ROC must be a connected area.



Calculation of inverse Z-transform

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

C is a closed curve

Example: $X(z) = \frac{1}{1 - 0.5z^{-1}}$

$$|z| > 0.5$$

$$X(z) = \frac{1}{1 - 0.5z^{-1}} = 1 + 0.5z^{-1} + 0.25z^{-2} + 0.125z^{-3} + \dots$$

$$x(n) = \begin{cases} 1, 0.5, 0.25, 0.125, \dots & n \geq 0 \\ 0 & n < 0 \end{cases} \Rightarrow x(n) = (0.5)^n u(n)$$



Inverse z-transform

- If the Z transform can be expanded out as a series in powers of z, then the coefficients of each term of the series constitutes the inverse. In the following expression, the inverse would be the coefficients in blue

$$X_0(z) = x_0(0) + x_0(1)z^{-1} + x_0(2)z^{-2} + x_0(3)z^{-3} + \dots$$

- Consider a Z transform which can be expanded as in the expression below

$$X(z) = 1 + \frac{1}{2}z^{-1} + \frac{1}{4}z^{-2} + \frac{1}{8}z^{-3} + \dots$$

- Then the inverse is the coefficients in blue and can be written as follows.

$$x(k) = 2^{-k}$$

Laplace, Fourier and z-Transforms



	Definition	Purpose	Suitable for ..
Laplace transform	$X(s) = \int_{-\infty}^{\infty} x(t)e^{-st} dt$	Converts integro-differential equations to algebraic equations	Continuous-time system & signal analysis; stable or unstable
Fourier transform	$X(\omega) = \int_{-\infty}^{\infty} x(t)e^{-j\omega t} dt$	Converts finite time signal to frequency domain	Continuous-time; stable system, convergent signals only; best for steady-state
Discrete Fourier transform	$X[n\omega_0] = \sum_{n=0}^{N_0-1} x[n]e^{jn\omega_0 T}$ <p>N_0 samples, T = sample period $\omega_0 = 2\pi / T$</p>	Converts finite discrete-time signal to discrete frequency domain	Discrete time, otherwise same as FT
z transform	$X[z] = \sum_{n=-\infty}^{\infty} x[n]z^{-n}$	Converts difference equations into algebraic equations	Discrete-time system & signal analysis; stable or unstable