

Systems

Systems map input (signal) to output (signal).

noisy image $\xrightarrow{\text{filter}}$ denoised image

trades $\xrightarrow{\text{market}}$ price

sensor \longrightarrow actuator



Notation: $y(t) = H(x(t))$ \leftarrow really a signal

System Properties

- Memoryless: H is memoryless if $y(t) = H(x(t))$ depends only on current input $x(t)$.

Eg: $y(t) = \sin(t) + \underline{x(t)}$ ✓

$y(t) = \sin(t-1) + \underline{x(t)}$ ✓

$y(t) = \int_0^t x(s) ds$ ✗

$y(t) = \int_{-t}^t x(s) ds$ ✗

- Causal: Say H is causal if $y(t)$ depends only on $x(s)$, $\forall s \leq t$.

If $x_1(t) = x_2(t)$, $\forall t \leq T$,

Then $y_1(t) = y_2(t)$, $\forall t \leq T$.

" " " "
 $H(x_1(t))$ $H(x_2(t))$

Note: Memoryless \Rightarrow Causal.

- Time-Invariance: \mathcal{H} is T.I. if $y(t) = \mathcal{H}(x(t)) \Rightarrow y(t-t_0) = \mathcal{H}(x(t-t_0)), \forall t_0.$

Eg: $y(t) = \sin(t) + x(t) \xrightarrow{\text{Time Shift}} \sin(t-t_0) + x(t-t_0)$

$x_{\text{NEW}}(t) = x(t-t_0)$
 $\Rightarrow y_{\text{NEW}}(t) = \sin(t) + x_{\text{NEW}}(t)$
 $= \sin(t) + x(t-t_0)$

$y(t) = x(t) + x(t+1)$ ✓

- Linear: Say \mathcal{H} is linear if $\left. \begin{matrix} x_1(t) \xrightarrow{\mathcal{H}} y_1(t) \\ x_2(t) \xrightarrow{\mathcal{H}} y_2(t) \end{matrix} \right\} \Rightarrow a \cdot x_1(t) + b \cdot x_2(t) \xrightarrow{\mathcal{H}} a \cdot y_1(t) + b \cdot y_2(t)$ for any constants $a, b.$

Eg: Fourier transform.

- Invertible: \mathcal{H} is invertible if you can recover $x(t)$ from $y(t) = \mathcal{H}(x(t)).$
 No loss of information. " \mathcal{H}^{-1} exists"

- Stable: \mathcal{H} is stable if $x(t) \leq B_x, \forall t$, then $y(t) = \mathcal{H}(x(t)) \leq B_y, \forall t.$
 Output cannot grow too large.

LTI Systems

Impulse Response

Fingerprint of System

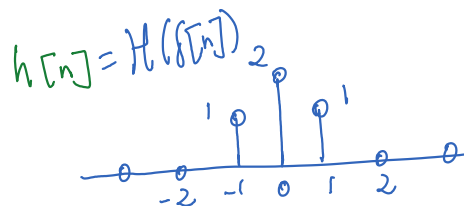
The impulse response of \mathcal{H} is

$h[n] = \mathcal{H}(\delta[n]).$

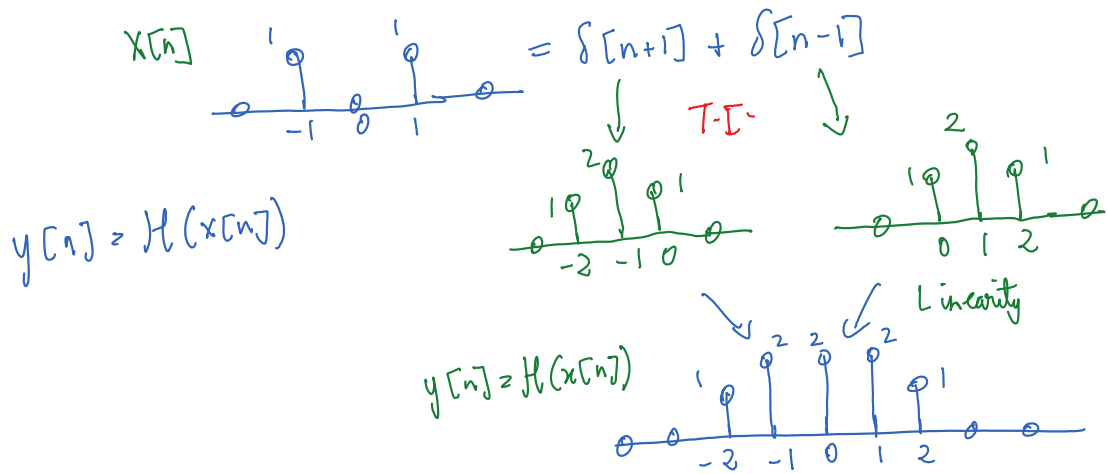
(Kronecker) Discrete delta function $\delta[n]$



Eg:



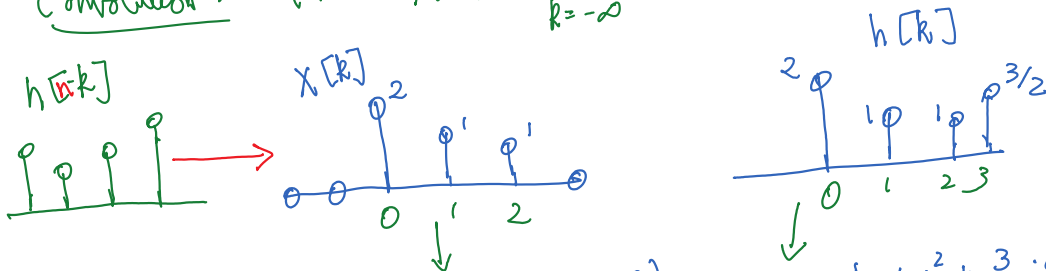
LTI System



For any LTI system,

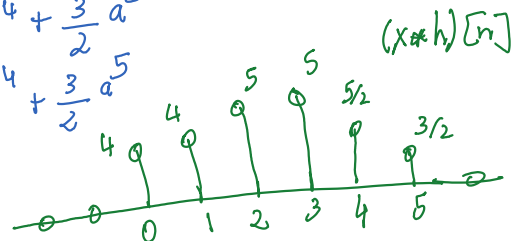
$$\begin{aligned}
 y[n] &= H(x[n]) \\
 &= H\left(\sum_{k=-\infty}^{\infty} x[k] \delta[n-k]\right) \\
 &= \sum_{k=-\infty}^{\infty} x[k] H(\delta[n-k]) \quad \text{Linearity} \\
 &= \sum_{k=-\infty}^{\infty} x[k] h[n-k] \quad \text{Time Invariance} \\
 &= (x * h)[n]
 \end{aligned}$$

Convolution: $(x * h)[n] = \sum_{k=-\infty}^{\infty} x[k] h[n-k]$



Polynomial Trick:

$$\begin{aligned}
 &(2 \cdot a^0 + 1 \cdot a^1 + 1 \cdot a^2) (2 \cdot a^0 + 1 \cdot a^1 + 1 \cdot a^2 + \frac{3}{2} \cdot a^3) \\
 &= 4 + 2a + 2a^2 + 3a^3 \\
 &\quad + 2a + a^2 + a^3 + \frac{3}{2}a^4 \\
 &\quad + 2a^2 + a^3 + a^4 + \frac{3}{2}a^5 \\
 &= 4 + 4a + 5a^2 + 5a^3 + \frac{5}{2}a^4 + \frac{3}{2}a^5
 \end{aligned}$$



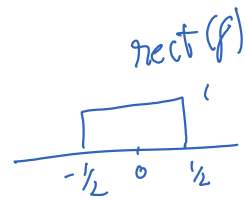
(AT) Fourier Transform and Convolution

$$F[y[n]] = \sum_{n=-\infty}^{\infty} y[n] e^{-i2\pi f n} \quad (\text{Analysis})$$

$$\begin{aligned}
 y[n] &= H(x[n]) \\
 &= \sum_{n=-\infty}^{\infty} \left(\sum_{k=-\infty}^{\infty} x[k] h[n-k] \right) e^{-i2\pi f k} e^{-i2\pi f (n-k)} \\
 &= \underbrace{\sum_{k=-\infty}^{\infty} x[k] e^{-i2\pi f k}}_{F[x[n]]} \underbrace{\sum_{n=-\infty}^{\infty} h[n-k] e^{-i2\pi f (n-k)}}_{F[h[n]]}
 \end{aligned}$$

Convolution Property

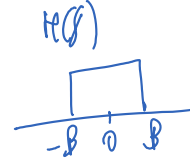
$$F[(x * h)[n]] = F[x[n]] \cdot F[h[n]].$$



Application: -) Efficient multiplication of polynomials

Reduction from N^2 operations to $N \cdot \log N$ operations.

-) Filtering



Equivalent to convolution with $\text{sinc}(ft)$.