

Notes on Discrete Time Signal Processing 3

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November 17, 2022

1 Discrete Time Signals

$$x[n] = x(nT) \quad (1) \quad \text{A discrete signal } x \text{ sampled at period } T.$$

1.1 Basic sequences

$$\delta[n] = \begin{cases} 0 & n \neq 0 \\ 1 & n = 0 \end{cases} \quad (2) \quad \text{This is the **unit sample** (impulse) function.}$$

$$x[n] = \sum_{k=-\infty}^{\infty} \delta[k-n]x[k] \quad (3) \quad \text{This derivation seems quite redundant, however it will be useful later on when analyzing linear time invariant systems.}$$

$$u[n] = \begin{cases} 0 & n < 0 \\ 1 & n \geq 0 \end{cases} = \sum_{k=-\infty}^{\infty} \delta = \sum_{k=0}^{\infty} \delta[n-k] \quad (4) \quad \text{This is the **unit step** function.}$$

$$x[n] = A\alpha^n \quad (5) \quad \text{General exponential sequence.}$$

$$x[n] = |A||\alpha|^n e^{j(\omega_0 n + \phi)} \quad (6) \quad \text{Setting } A = |A|e^{j\phi} \text{ and } \alpha = |\alpha|e^{j\omega_0}.$$

$$= |A||\alpha|^n [\cos(\omega_0 n + \phi)\sin(\omega_0 n + \phi)] \quad (7)$$

$$x[n] = |A||\alpha|^n e^{(\omega_1 n + \phi)j} \quad (8)$$

$$= |A||\alpha|^n e^{((2\pi + \omega_0)n + \phi)j} \quad (9)$$

$$= |A||\alpha|^n e^{(\omega_0 n + \phi)j} e^{2\pi nj} \quad (10)$$

$$= |A||\alpha|^n e^{(\omega_0 n + \phi)j} 1 \quad (11)$$

It could be noted that the effective frequency range in discrete systems is 2π because n is an integer. As an example, here we set $\omega_1 = 2\pi + \omega_0$.

2 Discrete time systems

$$y[n] = T\{x[n]\} \quad (12)$$

General representation of discrete time systems. Note that the output sequence at each value of the index n may depend on input samples $x[n]$ of all n .

$$\begin{aligned} T\{x_1[n] + x_2[n]\} &= T\{x_1[n]\} + T\{x_2[n]\} \\ &= y_1[n] + y_2[n] \end{aligned} \quad (13)$$

Linear systems are defined by the principle of superposition. Equation 13 follows the property of *additivity*, and Equation 14 follows the property of *scaling* or *homogeneity*.

$$T\{ax[n]\} = aT\{x[n]\} = ay[n] \quad (14)$$

$$y[n] = T\left\{\sum_{-\infty}^{\infty} x[k]\delta[n-k]\right\} \quad (15)$$

Linear time-invariant systems (LTI). Because of the time-invariant property, $h_k[n]$ in Equation 17 can be used for all values of k .

$$= \sum_{-\infty}^{\infty} x[k]T\{\delta[n-k]\} \quad (16)$$

$$= \sum_{-\infty}^{\infty} x[k]h_k[n] \quad (17)$$

$$= \sum_{-\infty}^{\infty} x[k]h[n-k] \quad (18)$$

$$y[n] = x[n] * h[n] \quad (19)$$

As a consequence of Equation 18, LTI systems can be represented as a convolutional operator on the input data.

$$x[n] * h[n] = h[n] * x[n] \quad (20)$$

Properties of LTI systems. LTI are commutative (Equation 20), distributive (Equation 21), associative (Equation 22). Therefore, multiple LTI systems can be combined into one.

$$x[n] * (h_1[n] + h_2[n]) = x[n] * h_1[n] + x[n] * h_2[n] \quad (21)$$

$$y[n] = (x[n] * h_1[n]) * h_2[n] = x[n] * (h_1[n] * h_2[n]) \quad (22)$$

$$h[n] = 0, \quad n < 0 \qquad (23) \quad \text{Causal LTI system only have response for } n \geq 0.$$
