Module 1 : Signals in Natural Domain Lecture 5 : Discrete-Time Convolution

Objectives

In this lecture you will learn the following

- We shall look into the properties of systems satisfying both linearity and shift invariance i.e. LSI (Linear shift invariant) systems.
- We shall define the term "Impulse response" in context to LSI systems.
- We shall learn Convolution, an operation which helps us find the output of the LTI system given the impulse response and the input signal.

[NOTE:- In the following sections we will be using LSI and LTI interchangeably.LTI is infact a special case of LSI. In LTI we consider shifts in time.]

Discrete time convolution

As the name suggests the two basic properties of a LTI system are:

1) Linearity

A linear system (continuous or discrete time) is a system that possesses the property of SUPERPOSITION. The principle of superposition states that the response of sum of two or more weighted inputs is the sum of the weighted responses of each of the signals. Mathematically

$$y[n] = S a_k y_k[n] = a_1 y_1[n] + a_2 y_2[n] +$$

Superposition combines in itself the properties of ADDITIVITY and HOMOGENEITY. This is a powerful property and allows us to evaluate the response for an arbitrary input, if it can be expressed as a sum of functions whose responses are known.

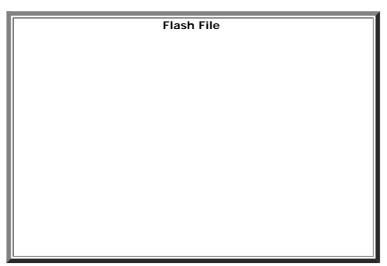
2) Time Invariance

It allows us to find the response to a function which is delayed or advanced in time; but similar in shape to a function whose response is known.

Given the response of a system to a particular input, these two properties enable us to find the response to all its delays or advances and their linear combination.

Discrete Time LTI Systems

Consider any discrete time signal **x[n]**. It is intuitive to see how the signal **x[n]** can be represented as sum of many delayed/advanced and scaled Unit Impulse Signals.



Mathematically, the above function can be represented as

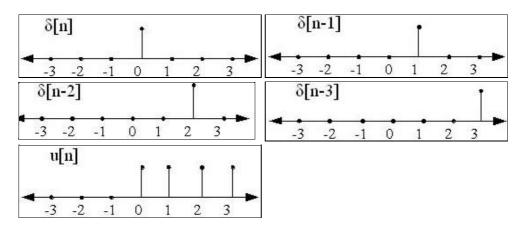
$$x[n] = x[-2] \delta[n+2] + x[-1] \delta[n+1] + x[0] \delta[n] + x[1] \delta[n-1] + x[2] \delta[n-2] + x[3] \delta[n-3]$$

More generally any discrete time signal x[n] can be represented as

$$x[n] = \sum_{k=-\infty}^{\infty} x[k] \delta[n-k]$$

The above expression corresponds to the representation of any arbitrary sequence as a linear combination of shifted Unit Impulses $\delta[n-k]$ which are scaled by $\mathbf{x}[\mathbf{n}]$. Consider for example the Unit Step function. As shown earlier it can be represented as

$$\mathbf{u}[\mathbf{n}] = \sum_{k=0}^{\infty} \delta[\mathbf{n} - \mathbf{k}]$$



Now if we knew the response of a system for a Unit Impulse Function, we can obtain the response of any arbitrary input. To see why this is so, we invoke the properties of Linearity, Homogeneity (Superposition) and Time Invariance.

Let h[n] be response of system to the unit impulse function \[\delta[n] \]

We represent this as $\delta[n] \longrightarrow h[n]$

Because of shift-invariance $\delta[n-k] \rightarrow h[n-k]$

Now invoking homogenity $x[k] \delta[n-k] \rightarrow x[k] h[n-k]$

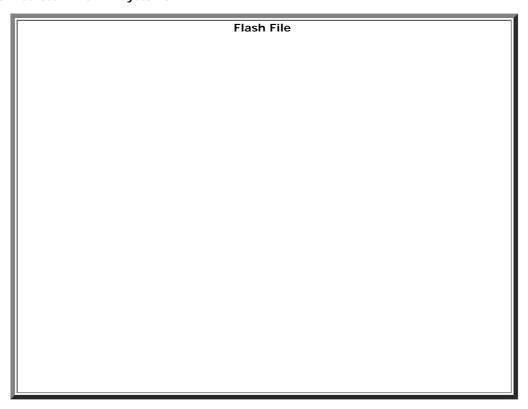
Because of additivity $\sum x[k] \, \delta[n-k] \longrightarrow \sum x[k] \, h[n-k]$ $-\omega \le k \le \omega$

The left hand side can be identified as any arbitrary input, while the right hand side can be identified as the total output to the signal. The total response of the system is referred to as the CONVOLUTION SUM or superposition sum of the sequences x[n] and h[n]. The result is more concisely stated as y[n] = x[n] * h[n], where

$$x[n] * h[n] = \sum x[k] h[n-k] -\infty \le k \le \infty$$

Therefore, as we said earlier a LTI system is completely characterized by its response to a single signal i.e. response to the Unit Impulse signal.

Example Related to Discrete Time LTI Systems

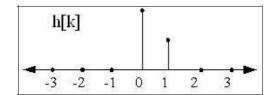


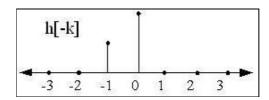
Recall that the convolution sum is given by

$$y[n] = \sum_{k=-\infty}^{\infty} x[k] h[n-k]$$

Now we plot x[k] and h[n-k] as functions of k and not n because of the summation over k. Functions x[k] and h[k] are the same as x[n] and h[n] but plotted as functions of k. Then, the convolution sum is realized as follows

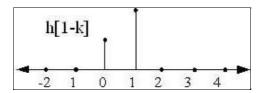
1. Invert h[k] about k=0 to obtain h[-k].



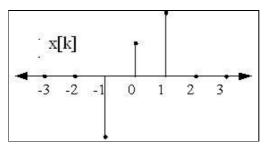


2. The function h[n-k] is given by h[-k] shifted to the right by n (if n is positive) and to the left (if n is negative). It may appear contradictory but think a while to verify this (note the sign of the independent variable).

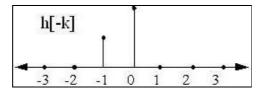
In the figure below n=1



3. Multiply $\mathbf{x}[\mathbf{k}]$ and $\mathbf{h}[\mathbf{n}-\mathbf{k}]$ for same coordinates on the k axis. The value obtained is the response at \mathbf{n} i.e. Value of $\mathbf{y}[\mathbf{n}]$ at a particular \mathbf{n} the value chosen in step 2. Now we demonstrate the entire procedure taking $\mathbf{n}=0,1$ thereby obtaining the response at $\mathbf{n}=0,1$. The input signal $\mathbf{x}[\mathbf{n}]$ and for this example is taken as :



Case 1: For n=0

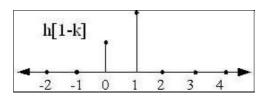


Remember the independence axis has \mathbf{k} as the independent variable. Then taking the product $\mathbf{x}[\mathbf{k}] \mathbf{h}[-\mathbf{k}]$ for same \mathbf{k} and summing it we get the value of the response at $\mathbf{n}=\mathbf{0}$.

Let
$$h[-k] = g[k]$$

$$y[0] =x[-1]g[-1] + x[0]g[0] + = (-2)(1) + (1)(2) = 0$$

Case 2: For n=1



$$h[1-k] = g[k]$$

$$y[1] = \dots + x[0]g[0] + x[1]g[1] + \dots = (1)(1) + (2)(2) = 5$$

The values are the same as that obtained previously.

The total response referred to as the Convolution sum need not always be found graphically. The formula can directly be applied if the input and the impulse response are some mathematical functions. We show this by an example next.

Example

 $x[n] = \left(\frac{1}{2}\right)^n u[n]$. And the impulse response is given by $h[n] = \left(\frac{1}{3}\right)^n u[n]$. Find the total response when the input function is

Applying the convolution formula we get

$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

$$= \sum_{k=-\infty}^{\infty} \left(\frac{1}{2}\right)^k u[k] \left(\frac{1}{3}\right)^{n-k} u[n-k]$$

$$= \sum_{k=0}^{n} \left(\frac{1}{2}\right)^{k} \left(\frac{1}{3}\right)^{n-k}$$
 because $u[k] = 0$ for $k < 0$ and $u[n-k] = 0$ for $k > n$

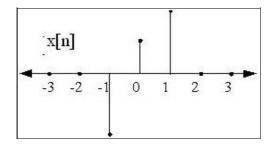
$$= \left(\frac{1}{3}\right)^n \sum_{k=0}^n \left(\frac{3}{2}\right)^k$$

$$= \left(\frac{1}{3}\right)^n \quad \frac{1 - \left(\frac{3}{2}\right)^{n+1}}{1 - \left(\frac{3}{2}\right)}$$

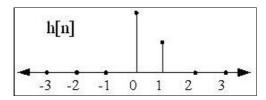
$$= (-2) \left(\frac{1}{3}\right)^n u[n] + 3 \left(\frac{1}{2}\right)^n u[n]$$

We now give an alternative method for calculating the convolution of the given signal x[n] and the response to the unit impulse function. Let us see how convolution output is the sum of weighted and shifted instances of the impulse response.

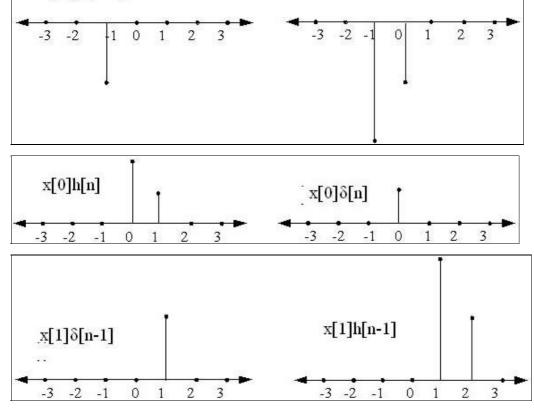
Let the given signal x[n] be



Let the Impulse Response be



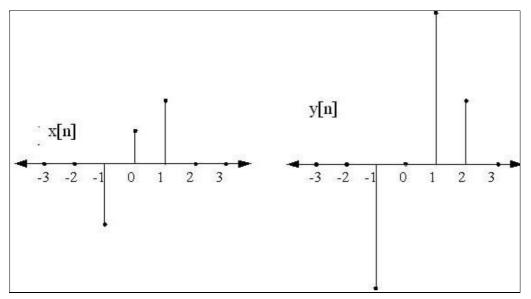
Now we break the signal in its components i.e. expressed as a sum of unit impulses scaled and delayed or advanced appropriately. Simultaneously we show the output as sum of responses of unit impulses function scaled by the same multiplying factor and appropriately delayed or advanced.



 $x[-1]\delta[n+1]$

Summing the left and the right hand sides of the above figures we get the input x[n] and the total response on the left and the right sides respectively. Thus we see the graphical analog the above formula.

x[-1]h[n+1]



The total response referred to as the Convolution sum need not always be found graphically. The formula can directly be applied if the input and the impulse response are some mathematical functions. We show this by a example.

Conclusion:

In this lecture you have learnt:

- The two basic properties of LTI systems are **linearity** and **shift-invariance**. It is completely characterised by its impulse response.
- Any discrete time signal x[n] can be represented as a linear combination of shifted Unit Impulses scaled by x[n].
- The unit step function can be represented as sum of shifted unit impulses.
- The total response of the system is referred to as the CONVOLUTION SUM or superposition sum of the sequences x[n] and h[n]. The result is more concisely stated as y[n] = x[n] * h[n].
- The convolution sum is realized as follows
 - 1. Invert h[k] about k=0 to obtain h[-k].
 - 2. The function **h[n-k]** is given by h[-k] shifted to the right by n (if n is positive) and to the left (if n is negative) (note the sign of the independent variable).
 - 3. Multiply $\mathbf{x}[k]$ and $\mathbf{h}[n-k]$ for same coordinates on the k axis. The value obtained is the response at n i.e. Value of y[n] at a particular n the value chosen in step 2.

Congratulations, you have finished Lecture 5.