

# Lecture 7: Time-Domain Analysis of Discrete-Time Systems

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- Discrete-Time System Equations
- $E$  operator
- Response of Linear Discrete-Time Systems
- Zero-Input and Zero-State Response
- Convolution Sum
- Graphical Procedure for the Convolution Sum
- Classical Solution of Linear Difference Equations

# Discrete-Time System Equations

## Advanced operator form:

$$y[k+n] + a_{n-1}y[k+n-1] + \cdots + a_1y[k+1] + a_0y[k] = \\ b_m f[k+m] + b_{m-1}f[k+m-1] + \cdots + b_1f[k+1] + b_0f[k]$$

- The left-hand side of this form consists of the output at instants  $k+n$ ,  $k+n-1$ ,  $k+n-2$ , and so on. The right-hand side of the equation consists of the input at instants  $k+m$ ,  $k+m-1$ ,  $k+m-2$ , and so on.
- The condition that the above system is causal is  $m \leq n$ . For a general causal case,  $m = n$ , the above system can be expressed as

$$y[k+n] + a_{n-1}y[k+n-1] + \cdots + a_1y[k+1] + a_0y[k] =$$

$$b_n f[k+n] + b_{n-1}f[k+n-1] + \cdots + b_1f[k+1] + b_0f[k]$$

# Discrete-Time System Equations

## Delay operator form:

In case  $m = n$ , we can replace  $k$  by  $k - n$  throughout the equation.

- Such replacement yields the delay operator form.

$$y[k] + a_{n-1}y[k-1] + \cdots + a_1y[k-n+1] + a_0y[k-n] = \\ b_nf[k] + b_{n-1}f[k-1] + \cdots + b_1f[k-n+1] + b_0f[k-n]$$

# Iterative Solution of Difference Equations

From the delay operator form

$$y[k] + a_{n-1}y[k-1] + \cdots + a_1y[k-n+1] + a_0y[k-n] = \\ b_nf[k] + b_{n-1}f[k-1] + \cdots + b_1f[k-n+1] + b_0f[k-n]$$

It can be expressed as

$$y[k] = -a_{n-1}y[k-1] - a_{n-2}y[k-2] - \cdots - a_0y[k-n] \\ + b_nf[k] + b_{n-1}f[k-1] + \cdots + b_0f[k-n]$$

There are the past  $n$  values of the output:  $y[k-1], y[k-2], \dots, y[k-n]$ , the past  $n$  values of the input:  $f[k-1], f[k-2], \dots, f[k-n]$ , and the present value of the input  $f[k]$ .

# Initial Conditions and Iterative Solution of Difference Equations

If the input is causal, the  $f[-1] = f[-2] = \dots = f[-n] = 0$ , and we need only  $n$  initial conditions  $y[-1], y[-2], \dots, y[-n]$ . This result allows us to compute iteratively or recursively the output  $y[0], y[1], y[2], y[3], \dots$ , and so on. For instant,

- to find  $y[0]$  we set  $k = 0$ .
- the left-hand side is  $y[0]$ , and the right-hand side contains terms  $y[-1], y[-2], \dots, y[-n]$  and the input  $f[0], f[-1], f[-2], \dots, f[-n]$ .
- Therefore, we must know the  $n$  initial conditions  $y[-1], y[-2], \dots, y[-n]$  to find  $y[0], y[1], y[2], \dots$  and so on.

# Iterative Solution of Difference Equations

## Examples

Solve iteratively

$$y[k] - 0.5y[k-1] = f[k]$$

with initial condition  $y[-1] = 16$  and causal input  $f[k] = k^2$ .

**Solution:**

Rewritten the equation in the delay operator form and move all past outputs to the left:

$$y[k] = 0.5y[k-1] + f[k]$$

We obtain

$$y[0] = 0.5y[-1] + f[0] = 0.5(16) + 0 = 8$$

$$y[1] = 0.5y[0] + f[1] = 0.5(8) + (1)^2 = 5$$

$$y[2] = 0.5y[1] + f[2] = 0.5(5) + (2)^2 = 6.5$$

$$y[3] = 0.5y[2] + f[3] = 0.5(6.5) + (3)^2 = 12.25$$

$$y[4] = 0.5y[3] + f[4] = 0.5(12.25) + (4)^2 = 22.125$$

# Iterative Solution of Difference Equations

Examples cont.

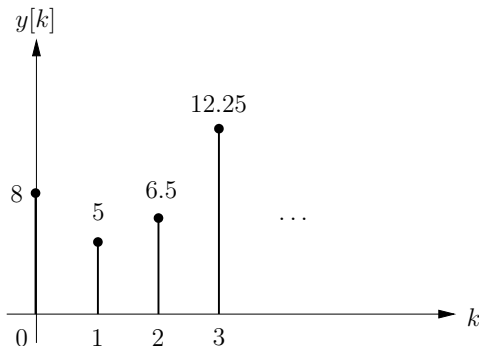


Figure: Iterative solution of a difference equation



# Iterative Solution of Difference Equations

## Examples cont.

Solve iteratively

$$y[k+2] - y[k+1] + 0.24y[k] = f[k+2] - 2f[k+1]$$

with initial conditions  $y[-1] = 2$ ,  $y[-2] = 1$  and a causal input  $f[k] = k$ .

**Solution:**

Rewritten the equation in the delay operator form and move all past outputs to the left:

$$y[k] = y[k-1] - 0.24y[k-2] + f[k] - 2f[k-1]$$

We obtain

$$y[0] = y[-1] - 0.24y[-2] + f[0] - 2f[-1] = 2 - 0.24(1) + 0 - 0 = 1.76$$

$$y[1] = y[0] - 0.24y[-1] + f[1] - 2f[0] = 1.76 - 0.24(2) + 1 - 0 = 2.28$$

$$y[2] = y[1] - 0.24y[0] + f[2] - 2f[1] = 2.28 - 0.24(1.76) + 2 - 2(1) = 1.8576$$

$$y[3] = y[2] - 0.24y[1] + f[3] - 2f[2] = 1.8576 - 0.24(2.28) + 3 - 2(2) = 0.3104$$

$$y[4] = y[3] - 0.24y[2] + f[4] - 2f[3] = 0.3104 - 0.24(1.8576) + 4 - 2(3) = -2.1354$$

# Iterative Solution of Difference Equations

Examples cont.

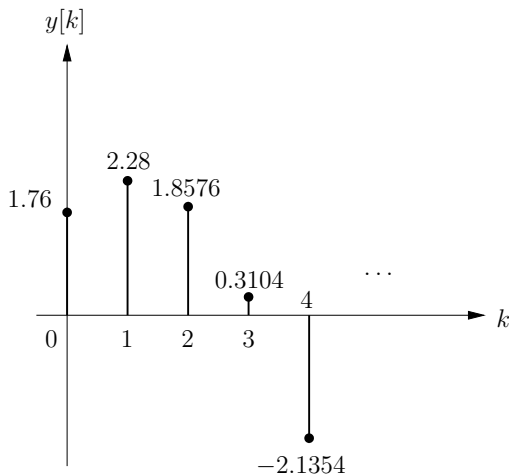


Figure: Iterative solution of a difference equation

# $E$ operator

In continuous-time system we used the operator  $D$  to denote the operation of differentiation. For discrete-time systems we use the operator  $E$  to denote the operation for advancing the sequence by one time unit. Thus

$$\begin{aligned}Ef[k] &= f[k + 1] \\E^2f[k] &= f[k + 2] \\&\vdots \\E^nf[k] &= f[k + n]\end{aligned}$$

For example

$$\begin{aligned}y[k + 1] - ay[k] &= f[k + 1] \\Ey[k] - ay[k] &= Ef[k] \\(E - a)y[k] &= Ef[k]\end{aligned}$$

# $E$ operator

cont.

For the second-order difference equation

$$y[k+2] + \frac{1}{4}y[k+1] + \frac{1}{16}y[k] = f[k+2]$$
$$\left(E^2 + \frac{1}{4}E + \frac{1}{16}\right)y[k] = E^2f[k]$$

A general  $n$ th-order difference equation ( $n = m$ ) can be expressed as

$$\begin{aligned} (E^n + a_{n-1}E^{n-1} + \dots + a_1E + a_0)y[k] = \\ (b_nE^n + b_{n-1}E^{n-1} + \dots + b_1E + b_0)f[k] \\ Q[E]y[k] = P[E]f[k] \end{aligned}$$

where

$$Q[E] = E^n + a_{n-1}E^{n-1} + \dots + a_1E + a_0$$
$$P[E] = b_nE^n + b_{n-1}E^{n-1} + \dots + b_1E + b_0$$

# Response of Linear Discrete-Time Systems

## System response to Internal Conditions: The Zero-Input Response

Similar to the continuous-time case,

$$\text{Total response} = \text{zero-input response} + \text{zero-state response}$$

The zero-input response  $y_0[k]$  is the solution of the system with  $f[k] = 0$ ; that is,

$$Q[E]y_0[k] = 0$$

or

$$(E^n + a_{n-1}E^{n-1} + \cdots + a_1E + a_0)y_0[k] = 0$$

$$y_0[k+n] + a_{n-1}y_0[k+n-1] + \cdots + a_1y_0[k+1] + a_0y_0[k] = 0$$

# Response of Linear Discrete-Time Systems

## System response to Internal Conditions: The Zero-Input Response cont.

The equation states that a linear combination of  $y_0[k]$  and advanced  $y_0[k]$  is zero not for some values of  $k$  but for all  $k$ . Such situation is possible if and only if  $y_0[k]$  and advanced  $y_0[k]$  have the same form. This is true only for an exponential function  $\gamma^k$ . Since

$$\gamma^{k+m} = \gamma^m \gamma^k$$

Therefore, if  $y_0[k] = c\gamma^k$  we have

$$\begin{aligned} Ey_0[k] &= y_0[k+1] = c\gamma^{k+1} = c\gamma\gamma^k \\ E^2y_0[k] &= y_0[k+2] = c\gamma^{k+2} = c\gamma^2\gamma^k \\ &\vdots \\ E^ny_0[k] &= y_0[k+n] = c\gamma^{k+n} = c\gamma^n\gamma^k \end{aligned}$$

# Response of Linear Discrete-Time Systems

System response to Internal Conditions: The Zero-Input Response cont.

Substitution of these results to the system equation yields

$$c(\gamma^n + a_{n-1}\gamma^{n-1} + \cdots + a_1\gamma + a_0)\gamma^k = 0$$

For a nontrivial solution of this equation

$$(\gamma^n + a_{n-1}\gamma^{n-1} + \cdots + a_1\gamma + a_0) = 0 \text{ or } Q[\gamma] = 0$$

$Q[\gamma]$  is an  $n$ th-order polynomial and can be expressed in the factorized form (assuming all distinct roots):

$$(\gamma - \gamma_1)(\gamma - \gamma_2) \cdots (\gamma - \gamma_n) = 0$$

Clearly,  $\gamma$  has  $n$  solutions  $\gamma_1, \gamma_2, \cdots, \gamma_n$  and, the system has  $n$  solutions  $c_1\gamma_1^k, c_2\gamma_2^k, \dots, c_n\gamma_n^k$ .

# Response of Linear Discrete-Time Systems

## System response to Internal Conditions: The Zero-Input Response cont.

The zero-input response is

$$y_0[k] = c_1\gamma_1^k + c_2\gamma_2^k + \cdots + c_n\gamma_n^k$$

where  $\gamma_1, \gamma_2, \dots, \gamma_n$  are the roots of the polynomial.

- $Q[\gamma]$  is called the **characteristic polynomial** of the system.
- $Q[\gamma] = 0$  is the **characteristic equation** of the system.
- $\gamma_1, \gamma_2, \dots, \gamma_n$  are called **characteristic roots** or **characteristic values** (also **eigenvalues**) of the system.
- The exponentials  $\gamma_i^k (i = 1, 2, \dots, n)$  are **characteristic modes** or **natural modes** of the system.



# Response of Linear Discrete-Time Systems

System response to Internal Conditions: The Zero-Input Response cont.

## Repeated Roots:

If two or more roots are repeated, the form of the characteristic modes is modified. Similar to the continuous-time case, if a root  $\gamma$  repeats  $r$  times, the characteristic modes corresponding to this root are  $\gamma^k$ ,  $k\gamma^k$ ,  $k^2\gamma^k$ ,  $\dots$ ,  $k^{r-1}\gamma^k$ .

If the characteristic equation of a system is

$$Q[\gamma] = (\gamma - \gamma_1)^r (\gamma - \gamma_{r+1}) (\gamma - \gamma_{r+2}) \cdots (\gamma - \gamma_n)$$

the zero-input response of the system is

$$y_0[k] = (c_1 + c_2 k + c_3 k^2 + \cdots + c_r k^{r-1}) \gamma_1^k \\ + c_{r+1} \gamma_{r+1}^k + c_{r+2} \gamma_{r+2}^k + \cdots + c_n \gamma_n^k$$

# Response of Linear Discrete-Time Systems

System response to Internal Conditions: The Zero-Input Response cont.

## Complex Roots:

As in the case of continuous-time systems, the complex roots of a discrete-time system must occur in pairs of conjugates so that the system equation coefficients are real. Like the case of continuous-time systems, we can eliminate dealing with complex numbers by using the real form of the solution.

- First express the complex conjugate roots  $\gamma$  and  $\gamma^*$  in polar form.

$$\gamma = |\gamma|e^{j\beta} \text{ and } \gamma^* = |\gamma|e^{-j\beta}$$

- the zero-input response is given by

$$\begin{aligned} y_0[k] &= C_1\gamma^k + C_2(\gamma^*)^k \\ &= C_1|\gamma|^k e^{j\beta k} + C_2|\gamma|^k e^{-j\beta k} \end{aligned}$$

# Response of Linear Discrete-Time Systems

## System response to Internal Conditions: The Zero-Input Response cont.

For a real system,  $C_1$  and  $C_2$  must be conjugates so that  $y_0[k]$  is a real function of  $k$ . Let

$$\begin{aligned}C_1 &= \frac{C}{2}e^{j\theta} \text{ and } C_2 = \frac{C}{2}e^{-j\theta} \\y_0[k] &= \frac{C}{2}|\gamma|^k \left[ e^{j(\beta k + \theta)} + e^{-j(\beta k + \theta)} \right] \\&= C|\gamma|^k \cos(\beta k + \theta)\end{aligned}$$

where  $C$  and  $\theta$  are arbitrary constants determined from the auxiliary conditions.

# Response of Linear Discrete-Time Systems

## System response to Internal Conditions: The Zero-Input Response cont.

For an LTID system described by the difference equation

$$y[k+2] - 0.6y[k+1] - 0.16y[k] = 5f[k+2]$$

Find the zero-input response  $y_0[k]$  of the system if the initial conditions are  $y[-1] = 0$  and  $y[-2] = \frac{25}{4}$ .

The system equation in  $E$  operator form is

$$(E^2 - 0.6E - 0.16)y[k] = 5E^2 f[k]$$

The characteristic equation is

$$\gamma^2 - 0.6\gamma - 0.16 = (\gamma + 0.2)(\gamma + 0.8) = 0$$

The zero-input response is

$$y_0[k] = C_1(-0.2)^k + C_2(0.8)^k$$

# Response of Linear Discrete-Time Systems

System response to Internal Conditions: The Zero-Input Response cont.

Substitute  $y_0[-1] = 0$  and  $y_0[-2] = \frac{25}{4}$  we obtain

$$-5C_1 + \frac{5}{4}C_2 = 0$$

$$25C_1 + \frac{25}{16}C_2 = \frac{25}{4}$$

and  $C_1 = \frac{1}{5}$  and  $C_2 = \frac{4}{5}$ . Therefore

$$y_0[k] = \frac{1}{5}(-0.2)^k + \frac{4}{5}(0.8)^k, \quad k \geq 0.$$

# Response of Linear Discrete-Time Systems

## System response to Internal Conditions: The Zero-Input Response cont.

A system specified by the equation

$$(E^2 + 6E + 9)y[k] = (2E^2 + 6E)f[k]$$

determine  $y_0[k]$ , the zero-input response, if the initial condition are  $y_0[-1] = -\frac{1}{3}$  and  $y_0[-2] = -\frac{2}{9}$ .

The characteristic equation is

$$\gamma^2 + 6\gamma + 9 = (\gamma + 3)^2 = 0$$

and we have a repeated characteristic root at  $\gamma = -3$ . Hence, the zero-input response is

$$y_0[k] = (C_1 + C_2 k)(-3)^k.$$

From the initial conditions we have

$$\begin{aligned}C_1 - C_2 &= 1 \\C_1 - 2C_2 &= -2\end{aligned}$$

and  $C_1 = 4$ ,  $C_2 = 3$ . Finally, we have  $y_0[k] = (4 + 3k)(-3)^k$ ,  $k \geq 0$ .

# Response of Linear Discrete-Time Systems

## System response to Internal Conditions: The Zero-Input Response cont.

Find the zero-input response of an LTID system described by the equation

$$(E^2 - 1.56E + 0.81)y[k] = (E + 3)f[k]$$

when the initial conditions are  $y_0[-1] = 2$  and  $y_0[-2] = 1$ .

The characteristic equation is

$$(\gamma^2 - 1.56\gamma + 0.81) = (\gamma - 0.78 - j0.45)(\gamma - 0.78 + j0.45) = 0.$$

The characteristic roots are  $0.78 \pm j0.45$ ; that is,  $0.9e^{\pm j\frac{\pi}{6}}$ . Thus,  $|\gamma| = 0.9$  and  $\beta = \frac{\pi}{6}$ , and the zero-input response is given by

$$y_0[k] = C(0.9)^k \cos\left(\frac{\pi}{6}k + \theta\right).$$

Substituting the initial conditions  $y_0[-1] = 2$  and  $y_0[-2] = 1$ , we obtain

$$\frac{C}{0.9} \cos\left(-\frac{\pi}{6} + \theta\right) = \frac{C}{0.9} \left[ \cos\left(-\frac{\pi}{6}\right) \cos \theta - \sin\left(-\frac{\pi}{6}\right) \sin \theta \right] = \frac{C}{0.9} \left[ \frac{\sqrt{3}}{2} \cos \theta + \frac{1}{2} \sin \theta \right] = 2$$

# Response of Linear Discrete-Time Systems

## System response to Internal Conditions: The Zero-Input Response cont.

and

$$\frac{C}{(0.9)^2} \cos\left(-\frac{\pi}{3} + \theta\right) = \frac{C}{0.81} \left[ \cos\left(-\frac{\pi}{3}\right) \cos \theta - \sin\left(-\frac{\pi}{3}\right) \sin \theta \right] = \frac{C}{0.81} \left[ \frac{1}{2} \cos \theta + \frac{\sqrt{3}}{2} \sin \theta \right] = 1$$

or

$$\begin{aligned} \frac{\sqrt{3}}{1.8} C \cos \theta + \frac{1}{1.8} C \sin \theta &= 2 \\ \frac{1}{1.62} C \cos \theta + \frac{\sqrt{3}}{1.62} C \sin \theta &= 1. \end{aligned}$$

We have  $C \cos \theta = 2.308$  and  $C \sin \theta = -0.397$ . Then

$$\theta = \tan^{-1} \frac{-0.397}{2.308} = -0.17 \text{ rad}$$

Substituting  $\theta = -0.17$  radian in  $C \cos \theta = 2.308$  yields  $C = 2.34$  and

$$y_0[k] = 2.34(0.9)^k \cos\left(\frac{\pi}{6}k - 0.17\right), \quad k \geq 0$$



# The Unit Impulse Response $h[k]$

Consider an  $n$ th-order system specified by the equation

$$\begin{aligned} (E^n + a_{n-1}E^{n-1} + \cdots + a_1E + a_0) y[k] = \\ (b_nE^n + b_{n-1}E^{n-1} + \cdots + b_1E + b_0) f[k] \end{aligned}$$

or

$$Q[E]y[k] = P[E]f[k]$$

The unit impulse response  $h[k]$  is the solution of this equation for the input  $\delta[k]$  with all the initial conditions zero; that is

$$Q[E]h[k] = P[E]\delta[k]$$

subject to initial conditions

$$h[-1] = h[-2] = \cdots = h[-n] = 0$$

# The Unit Impulse Response $h[k]$

$h[k]$  is the system response to the input  $\delta[k]$ , which is zero for  $k > 0$ .

We know that when the input is zero, only the characteristic modes can be sustained by the system. Therefore,  $h[k]$  must be made up of characteristic modes for  $k > 0$ . At  $k = 0$ , it may have some nonzero value, and  $h[k]$  can be expressed as

$$h[k] = \frac{b_0}{a_0} \delta[k] + y_n[k] u[k].$$

The  $n$  unknown coefficients in  $y_n[k]$  can be determined from a knowledge of  $n$  values of  $h[k]$ . It is a straightforward task to determine values of  $h[k]$  iteratively.

# The Unit Impulse Response $h[k]$

## The Closed-Form Solution of $h[k]$ Deviation

For a discrete-time system specified above, we have

$$h[k] = A_0\delta[k] + y_n[k]u[k]$$

Then

$$Q[E](A_0\delta[k] + y_n[k]u[k]) = P[E]\delta[k]$$

because  $y_n[k]u[k]$  is a sum of characteristic modes

$$Q[E](y_n[k]u[k]) = 0, \quad k \geq 0$$

The above equation reduces to

$$A_0Q[E]\delta[k] = P[E]\delta[k], \quad k \geq 0$$

# The Unit Impulse Response $h[k]$

The Closed-Form Solution of  $h[k]$  Deviation cont.

or

$$\begin{aligned} A_0 (E^n + a_{n-1}E^{n-1} + \cdots + a_1E + a_0) \delta[k] \\ &= (b_nE^n + b_{n-1}E^{n-1} + \cdots + b_1E + b_0) \delta[k] \\ A_0(\delta[k+n] + a_{n-1}\delta[k+n-1] + \cdots + a_1\delta[k+1] + a_0\delta[k]) \\ &= b_n\delta[k+n] + b_{n-1}\delta[k+n-1] + \cdots + b_1\delta[k+1] + b_0\delta[k] \end{aligned}$$

If we set  $k = 0$  in the equation and recognize that  $\delta[0] = 1$  and  $\delta[m] = 0$  when  $m \neq 0$ , all but the last terms vanish on both sides, yielding

$$A_0 a_0 = b_0 \quad \text{and} \quad A_0 = \frac{b_0}{a_0}$$

Note: for the special case  $a_0 = 0$  see the reference.

# The Unit Impulse Response $h[k]$

## Example

Determine the unit impulse response  $h[k]$  for a system specified by the equation

$$y[k] - 0.6y[k-1] - 0.16y[k-2] = 5f[k]$$

This equation can be expressed in the advance operator form as

$$y[k+2] - 0.6y[k+1] - 0.16y[k] = 5f[k+2]$$

or

$$(E^2 - 0.6E - 0.16)y[k] = 5E^2f[k]$$

The characteristic equation is

$$\gamma^2 - 0.6\gamma - 0.16 = (\gamma + 0.2)(\gamma - 0.8) = 0.$$

Therefore

$$y_n[k] = C_1(-0.2)^k + C_2(0.8)^k$$

# The Unit Impulse Response $h[k]$

Example cont.

From the system we have  $a_0 = -0.16$  and  $b_0 = 0$ . Therefore

$$h[k] = \frac{0}{-0.16} \delta[k] + \left[ C_1(-0.2)^k + C_2(0.8)^k \right] u[k] = \left[ C_1(-0.2)^k + C_2(0.8)^k \right] u[k]$$

To determine  $C_1$  and  $C_2$ , we need to find two values of  $h[k]$  iteratively. To do this, we must let the input  $f[k] = \delta[k]$  and the output  $y[k] = h[k]$  in the system equation. The resulting equation is

$$h[k] - 0.6h[k-1] - 0.16h[k-2] = 5\delta[k]$$

subject to zero initial state; that is,  $h[-1] = h[-2] = 0$ .

Setting  $k = 0$  in this equation yields

$$h[0] - 0.6(0) - 0.16(0) = 5(1) \implies h[0] = 5$$

Next, setting  $k = 1$  and using  $h[0] = 5$ , we obtain

# The Unit Impulse Response $h[k]$

Example cont.

Then we have

$$h[0] = C_1(-0.2)^0 + C_2(0.8)^0 = C_1 + C_2 = 5$$

$$h[1] = C_1(-0.2)^1 + C_2(0.8)^1 = -0.2C_1 + 0.8C_2 = 3$$

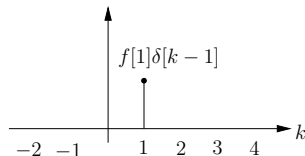
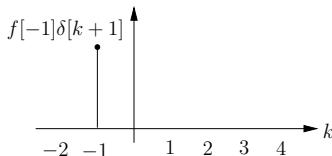
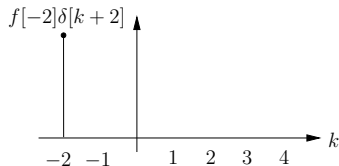
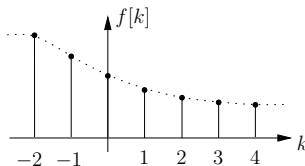
and  $C_1 = 1$ ,  $C_2 = 4$ . Therefore

$$h[k] = \left[ (-0.2)^k + 4(0.8)^k \right] u[k]$$

# System Response to External Input

## The Zero-State Response

The zero-state response  $y[k]$  is the system response to an input  $f[k]$  when the system is in zero state. Like in the continuous-time case an arbitrary input  $f[k]$  can be expressed as a sum of impulse components.





# System Response to External Input

## The Zero-State Response cont.

The previous page shows how a signal  $f[k]$  can be expressed as a sum of impulse components. The component of  $f[k]$  at  $k = m$  is  $f[m]\delta[k - m]$ , and  $f[k]$  is the sum of all these components summed from  $m = -\infty$  to  $\infty$ . Therefore

$$\begin{aligned} f[k] &= f[0]\delta[k] + f[1]\delta[k - 1] + f[2]\delta[k - 2] + \cdots \\ &\quad + f[-1]\delta[k + 1] + f[-2]\delta[k + 2] + \cdots \\ &= \sum_{m=-\infty}^{\infty} f[m]\delta[k - m] \end{aligned}$$

If we knew the system response to impulse  $\delta[k]$ , the system response to any arbitrary input could be obtained by summing the system response to various impulse components

# System Response to External Input

## The Zero-State Response cont.

If

$$\delta[k] \implies h[k]$$

then

$$\begin{aligned} \delta[k - m] &\implies h[k - m] \\ f[m]\delta[k - m] &\implies f[m]h[k - m] \\ \underbrace{\sum_{m=-\infty}^{\infty} f[m]\delta[k - m]}_{f[k]} &\implies \underbrace{\sum_{m=-\infty}^{\infty} f[m]h[k - m]}_{y[k]} \end{aligned}$$

# System Response to External Input

## The Zero-State Response cont.

We have the response  $y[k]$  to input  $f[k]$  as

$$y[k] = \sum_{m=-\infty}^{\infty} f[m]h[k-m].$$

This summation on the right-hand side is known as the **convolution sum** of  $f[k]$  and  $h[k]$ , and is represented symbolically by  $f[k] * h[k]$

$$f[k] * h[k] = \sum_{m=-\infty}^{\infty} f[m]h[k-m]$$

# Properties of the Convolution Sum

## The Commutative Property

### The Commutative Property

$$f_1[k] * f_2[k] = f_2[k] * f_1[k]$$

This can be proved as follow:

$$\begin{aligned} f_1[k] * f_2[k] &= \sum_{m=-\infty}^{\infty} f_1[m] f_2[k-m] \\ &= - \sum_{w=\infty}^{-\infty} f_1[w-k] f_2[w], \quad w = k - m \\ &= \sum_{w=-\infty}^{\infty} f_2[w] f_1[w-k] \\ &= f_2[k] * f_1[k] \end{aligned}$$

# Properties of the Convolution Sum

## The Distributive Property

### The Distributive Property

$$f_1[k] * (f_2[k] + f_3[k]) = f_1[k] * f_2[k] + f_1[k] * f_3[k]$$

The proof is as follow:

$$\begin{aligned} f_1[k] * (f_2[k] + f_3[k]) &= \sum_{m=-\infty}^{\infty} f_1[m] (f_2[k-m] + f_3[k-m]) \\ &= \sum_{m=-\infty}^{\infty} f_1[m] f_2[k-m] + \sum_{m=-\infty}^{\infty} f_1[m] f_3[k-m] \\ &= f_1[k] * f_2[k] + f_1[k] * f_3[k] \end{aligned}$$

# Properties of the Convolution Sum

## The Associative Property

### The Associative Property

$$f_1[k] * (f_2[k] * f_3[k]) = (f_1[k] * f_2[k]) * f_3[k]$$

The proof is as follow:

$$\begin{aligned} f_1[k] * (f_2[k] * f_3[k]) &= \sum_{m_1=-\infty}^{\infty} f_1[m_1] (f_2[k - m_1] * f_3[k - m_1]) \\ &= \sum_{m_1=-\infty}^{\infty} f_1[m_1] \sum_{m_2=-\infty}^{\infty} f_2[m_2] f_3[k - m_1 - m_2] \\ &= \sum_{m_1=-\infty}^{\infty} \sum_{m_2=-\infty}^{\infty} f_1[\lambda - m_2] f_2[m_2] f_3[k - \lambda] \end{aligned}$$

,where  $\lambda = m_1 + m_2$ .

# Properties of the Convolution Sum

## The Associative Property cont.

Then we have

$$\begin{aligned} f_1[k] * (f_2[k] * f_3[k]) &= \sum_{m_1=-\infty}^{\infty} \sum_{m_2=-\infty}^{\infty} f_1[\lambda - m_2] f_2[m_2] f_3[k - \lambda] \\ &= (f_1[k] * f_2[k]) * f_3[k] \end{aligned}$$

## The Convolution with an Impulse

$$f[k] * \delta[k] = \sum_{m=-\infty}^{\infty} f[m] \delta[k - m]$$

Since  $\delta[k - m] = 1$ , if  $k - m = 0$  or  $m = k$ , then

$$f[k] * \delta[k] = f[k].$$

# Properties of the Convolution Sum

## The shifting Property

### The shifting Property

If

$$f_1[k] * f_2[k] = c[k]$$

then

$$\begin{aligned} f_1[k] * f_2[k - n] &= f_1[k] * f_2[k] * \delta[k - n] \\ &= c[k] * \delta[k - n] = c[k - n] \end{aligned}$$

$$\begin{aligned} f_1[k - n] * f_2[k] &= f_1[k] * \delta[k - n] * f_2[k] \\ &= f_1[k] * f_2[k] * \delta[k - n] \\ &= c[k] * \delta[k - n] = c[k - n] \end{aligned}$$

$$\begin{aligned} f_1[k - n] * f_2[k - l] &= f_1[k] * \delta[k - n] * f_2[k] * \delta[k - l] \\ &= c[k] * \delta[k - n] * \delta[k - l] = c[k - n - l] \end{aligned}$$



# Properties of the Convolution Sum

## The shifting Property

### **The Width Property**

If  $f_1[k]$  and  $f_2[k]$  have lengths of  $m$  and  $n$  elements respectively, then the length of  $c[k]$  is  $m + n - 1$  elements.

# Causality and Zero-State Response

- We assumed the system to be linear and time-invariant.
- In practice, almost all of the input signals are causal, and a majority of the system are also causal.
- If the input  $f[k]$  is causal, then  $f[m] = 0$  for  $m < 0$ .
- Similarly, if the system is causal, then  $h[x] = 0$  for negative  $x$ , so that  $h[k - m] = 0$  when  $m > k$ .
- Therefore, if  $f[k]$  and  $h[k]$  are both causal, the product  $f[m]h[k - m] = 0$  for  $m < 0$  and for  $m > k$ , and it is nonzero only for the range  $0 \leq m \leq k$ . Therefore, the convolution sum is reduced to

$$y[k] = \sum_{m=0}^k f[m]h[k - m]$$

# Convolution Sum

## Analytical Method Example

Determine  $c[k] = f[k] * g[k]$  for

$$f[k] = (0.8)^k u[k] \quad \text{and} \quad g[k] = (0.3)^k u[k]$$

we have

$$c[k] = \sum_{m=0}^k f[m]g[k-m]$$

since both signals are causal.

$$c[k] = \begin{cases} \sum_{m=0}^k (0.8)^m (0.3)^{k-m} & k \geq 0 \\ 0 & k < 0 \end{cases}$$

$$\begin{aligned} c[k] &= (0.3)^k \sum_{m=0}^k \left(\frac{0.8}{0.3}\right)^m u[k] = (0.3)^k \frac{(0.8)^{k+1} - (0.3)^{k+1}}{(0.3)^k (0.8 - 0.3)} u[k] \\ &= 2 \left[ (0.8)^{k+1} - (0.3)^{k+1} \right] u[k] \end{aligned}$$

# Zero-State Response

## Analytical Method Example

Find the zero-state response  $y[k]$  of an LTID system described by the equation

$$y[k+2] - 0.6y[k+1] - 0.16y[k] = 5f[k+2]$$

if the input  $f[k] = 4^{-k}u[k]$  and  $h[k] = [(-0.2)^k + 4(0.8)^k]u[k]$ .

We have

$$\begin{aligned}y_s[k] &= f[k] * h[k] \\&= (4)^{-k}u[k] * [(-0.2)^k u[k] + 4(0.8)^k u[k]] \\&= (4)^{-k}u[k] * (-0.2)^k u[k] + (4)^{-k}u[k] * 4(0.8)^k u[k] \\&= (0.25)^k u[k] * (-0.2)^k u[k] + 4(0.25)^k u[k] * (0.8)^k u[k]\end{aligned}$$

Using Pair 4 from the convolution sum table:

$$y[k] = \left[ \frac{(0.25)^{k+1} - (-0.2)^{k+1}}{0.25 - (-0.2)} + 4 \frac{(0.25)^{k+1} - (0.8)^{k+1}}{0.25 - 0.8} \right] u[k]$$

# Zero-State Response

## Analytical Method Example cont.

$$\begin{aligned}y[k] &= \left( 2.22 \left[ (0.25)^{k+1} - (-0.2)^{k+1} \right] - 7.27 \left[ (0.25)^{k+1} - (0.8)^{k+1} \right] \right) u[k] \\&= \left[ -5.05(0.25)^{k+1} - 2.22(-0.2)^{k+1} + 7.27(0.8)^{k+1} \right] u[k]\end{aligned}$$

Recognizing that

$$\gamma^{k+1} = \gamma(\gamma)^k$$

We can express  $y[k]$  as

$$\begin{aligned}y[k] &= \left[ -1.26(0.25)^k + 0.444(-0.2)^k + 5.81(0.8)^k \right] u[k] \\&= \left[ -1.26(4)^{-k} + 0.444(-0.2)^k + 5.81(0.8)^k \right] u[k]\end{aligned}$$

# Graphical Procedure for the Convolution Sum

The convolution sum of causal signals  $f[k]$  and  $g[k]$  is given by

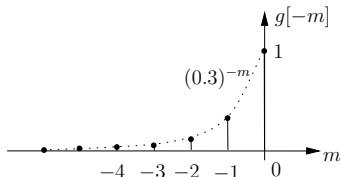
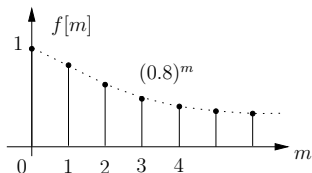
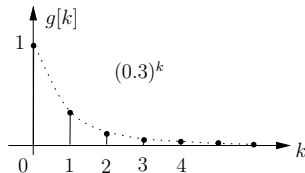
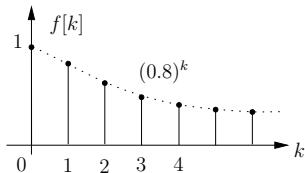
$$c[k] = \sum_{m=0}^k f[m]g[k-m]$$

- Invert  $g[m]$  about the vertical axis ( $m = 0$ ) to obtain  $g[-m]$ .
- Time shift  $g[-m]$  by  $k$  units to obtain  $g[k-m]$ . For  $k > 0$ , the shift is to the right (delay); for  $k < 0$ , the shift is to the left (advance).
- Next we multiply  $f[m]$  and  $g[k-m]$  and add all the products to obtain  $c[k]$ . The procedure is repeated to each value of  $k$  over the range  $-\infty$  to  $\infty$ .

# Graphical Procedure for the Convolution Sum

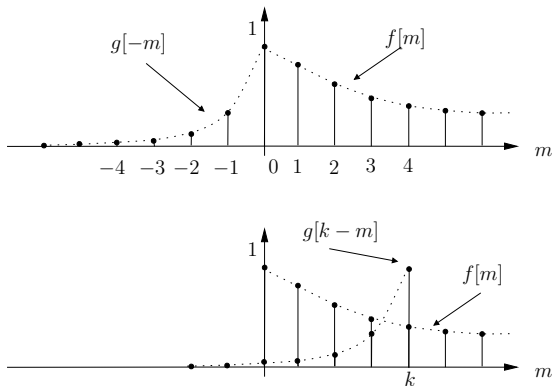
## Example

Find  $c[k] = f[k] * g[k]$ , where  $f[k]$  and  $g[k]$  are depicted in the Figures.



# Graphical Procedure for the Convolution Sum

## Example



The two functions  $f[m]$  and  $g[k-m]$  overlap over the interval  $0 \leq m \leq k$ .



# Graphical Procedure for the Convolution Sum

## Example

Therefore

$$\begin{aligned}c[k] &= \sum_{m=0}^k f[m]g[k-m] \\&= \sum_{m=0}^k (0.8)^m (0.3)^{k-m} \\&= (0.3)^k \sum_{m=0}^k \left(\frac{0.8}{0.3}\right)^m \\&= 2 \left[ (0.8)^{k+1} - (0.3)^{k+1} \right], \quad k \geq 0\end{aligned}$$

For  $k < 0$ , there is no overlap between  $f[m]$  and  $g[k-m]$ , so that  $c[k] = 0$   $k < 0$  and

$$c[k] = 2 \left[ (0.8)^{k+1} - (0.3)^{k+1} \right] u[k].$$

# Graphical Procedure for the Convolution Sum

## Sliding Tape Method

Using the sliding tape method, convolve the two sequences  $f[k]$  and  $g[k]$ .

- write the sequences  $f[k]$  and  $g[k]$  in the slots of two tapes
- leave the  $f$  tape stationary (to correspond to  $f[m]$ ). The  $g[-m]$  tape is obtained by time inverting the  $g[m]$
- shift the inverted tape by  $k$  slots, multiply values on two tapes in adjacent slots, and add all the products to find  $c[k]$ .

# Graphical Procedure for the Convolution Sum

## Sliding Tape Method cont.

For the case of  $k = 0$ ,

$$c[0] = 0 \times 1 = 0$$

For  $k = 1$

$$c[1] = (0 \times 1) + (1 \times 1) = 1$$

Similarly,

$$c[2] = (0 \times 1) + (1 \times 1) + (2 \times 1) = 3$$

$$c[3] = (0 \times 1) + (1 \times 1) + (2 \times 1) + (3 \times 1) = 6$$

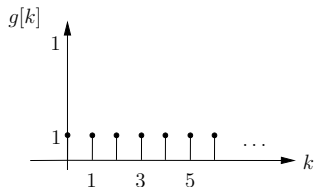
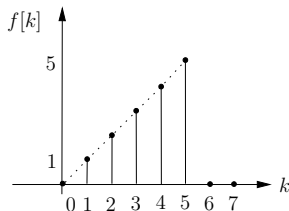
$$c[4] = (0 \times 1) + (1 \times 1) + (2 \times 1) + (3 \times 1) + (4 \times 1) = 10$$

$$c[5] = (0 \times 1) + (1 \times 1) + (2 \times 1) + (3 \times 1) + (4 \times 1) + (5 \times 1) = 15$$

$$c[6] = (0 \times 1) + (1 \times 1) + (2 \times 1) + (3 \times 1) + (4 \times 1) + (5 \times 1) = 15$$

# Graphical Procedure for the Convolution Sum

Sliding Tape Method cont.



0	1	2	3	4	5	
1	1	1	1	1	1	...

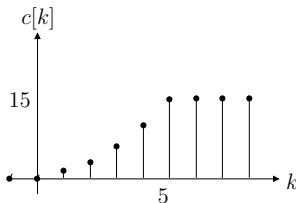
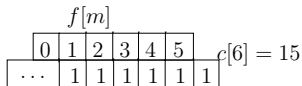
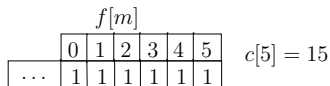
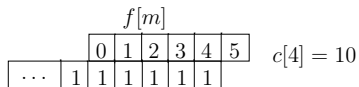
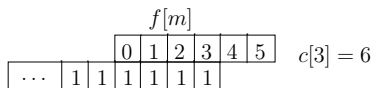
						$\leftarrow g[-m]$	0	1	2	3	4	5	$c[0] = 0$
...	1	1	1	1	1	1							

						$f[m]$						
						0	1	2	3	4	5	$c[1] = 1$
...	1	1	1	1	1	1	1					

						$f[m]$						
						0	1	2	3	4	5	$c[2] = 3$
...	1	1	1	1	1	1	1	1				

# Graphical Procedure for the Convolution Sum

Sliding Tape Method cont.



# Total Response

The total response of an LTID system can be expressed as a sum of the zero-input and zero-state components:

$$\text{Total response } y[k] = \underbrace{\sum_{j=1}^n c_j \gamma_j^k}_{\text{Zero-input component}} + \underbrace{f[k] * h[k]}_{\text{Zero-state component}}$$

From the previous example, the system described by the equation

$$y[k+2] - 0.6y[k+1] - 0.16y[k] = 5f[k+2]$$

with initial conditions  $y[-1] = 0$ ,  $y[-2] = \frac{25}{4}$  and input  $f[k] = (4)^{-k}u[k]$ . We have

$$y[k] = \underbrace{0.2(-0.2)^k + 0.8(0.8)^k}_{\text{Zero-input component}} \underbrace{-1.26(4)^{-k} + 0.444(-0.2)^k + 5.81(0.8)^k}_{\text{Zero-state component}}$$

# Classical solution of Linear Difference Equations

If  $y_n[k]$  and  $y_\phi[k]$  denote the natural and the forced response respectively, the the total response is given by

$$\text{Total response } y[k] = \underbrace{y_n[k]}_{\text{modes}} + \underbrace{y_\phi[k]}_{\text{nonmodes}}$$

Because  $y_n[k] + y_\phi[k]$  is a solution of the system, we have

$$Q[E](y_n[k] + y_\phi[k]) = P[E]f[k]$$

$y_n[k]$  is made up of characteristic modes,

$$Q[E]y_n[k] = 0$$

Substitution of this equation yields

$$Q[E]y_\phi[k] = P[E]f[k]$$

# Classical solution of Linear Difference Equations

## Forced Response

By definition, the forced response contains only nonmode terms and the list of the inputs and the corresponding forms of the forced function is show below:

Input $f[k]$	Forced Response $y_\phi[k]$
1. $r^k, r \neq \gamma_i \ (i = 1, 2, \dots, n)$	$cr^k$
2. $r^k, r = \gamma_i$	$ckr^k$
3. $\cos(\beta k + \theta)$	$c \cos(\beta k + \phi)$
4. $\left( \sum_{i=0}^m \alpha_i k^i \right) r^k$	$\left( \sum_{i=0}^m c_i k^i \right) r^k$

Note: By definition  $y_\phi[k]$  cannot have any characteristic mode terms.



# Classical solution of Linear Difference Equations

## Forced Response example

Determine the total response  $y[k]$  of a system

$$(E^2 - 5E + 6)y[k] = (E - 5)f[k]$$

if the input  $f[k] = (3k + 5)u[k]$  and the auxiliary conditions are  $y[0] = 4, y[1] = 13$ .  
The characteristic equation is

$$\gamma^2 - 5\gamma + 6 = (\gamma - 2)(\gamma - 3) = 0$$

Therefore, the natural response is

$$y_n[k] = B_1(2)^k + B_2(3)^k$$

To find the form of forced response  $y_\phi[k]$ , we use above Table, Pair 4 with  $r = 1, m = 1$ .  
This yields

$$y_\phi[k] = c_1k + c_0$$

# Classical solution of Linear Difference Equations

Forced Response example cont.

Therefore

$$y_\phi[k+1] = c_1(k+1) + c_0 = c_1k + c_1 + c_0$$

$$y_\phi[k+2] = c_1(k+2) + c_0 = c_1k + 2c_1 + c_0$$

Also

$$f[k] = 3k + 5$$

and

$$f[k+1] = 3(k+1) + 5 = 3k + 8$$

Substitution of the above results yields

$$c_1k + 2c_1 + c_0 - 5(c_1k + c_1 + c_0) + 6(c_1k + c_0) = 3k + 8 - 5(3k + 5)$$

$$2c_1k - 3c_1 + 2c_0 = -12k - 17$$

# Classical solution of Linear Difference Equations

## Forced Response example cont.

Comparison of similar terms on the two sides yields

$$2c_1 = -12$$

$$-3c_1 + 2c_0 = -17$$

and  $c_1 = -6$ ,  $c_0 = -\frac{35}{2}$ . Therefore

$$y_\phi[k] = -6k - \frac{35}{2}$$

The total response is

$$\begin{aligned} y[k] &= y_n[k] + y_\phi[k] \\ &= B_1(2)^k + B_2(3)^k - 6k - \frac{35}{2}, \quad k \geq 0 \end{aligned}$$

# Classical solution of Linear Difference Equations

## Forced Response example cont.

To determine arbitrary constants  $B_1$  and  $B_2$  we set  $k = 0$  and  $1$  and substitute the initial conditions  $y[0] = 4$ ,  $y[1] = 13$  to obtain

$$\begin{aligned} B_1 + B_2 - \frac{35}{2} &= 4 \\ 2B_1 + 3B_2 - \frac{47}{2} &= 13 \end{aligned}$$

and  $B_1 = 28$ ,  $B_2 = -\frac{13}{2}$ .  
Therefore

$$y_n[k] = 28(2)^k - \frac{13}{2}(3)^k$$

and

$$y[k] = \underbrace{28(2)^k - \frac{13}{2}(3)^k}_{y_n[k]} - \underbrace{6k - \frac{35}{2}}_{y_\phi[k]}$$

1. Lathi, B. P., *Signal Processing & Linear Systems*, Berkeley-Cambridge Press, 1998.