

# ECE 707: Control Systems Design (3)

T. Kirubarajan

Department of Electrical and Computer Engineering  
McMaster University  
Hamilton, Ontario, Canada

These viewgraphs are based on the text  
“Linear System: Theory and Design” by Chi-Tsong Chen  
Oxford University Press, 1999.

### Laplace Transform

Given the real function  $f(t)$  that satisfies the condition

$$\int_0^{\infty} |f(t)e^{-\sigma t}| dt < \infty \quad (1)$$

for some finite real  $\sigma$ , the Laplace transform of  $f(t)$  ( $L[f(t)]$ ) is defined as

$$\hat{f}(s) = \int_0^{\infty} f(t)e^{-st} dt \quad (2)$$

The variable  $s$  is complex variable, i.e.,  $s = \sigma + j\omega$ .

This particular definition (2) is known as **one-sided Laplace Transform**.

**Example 1:**  $f(t) = \begin{cases} 1 & t \geq 0 \\ 0 & t < 0 \end{cases}$ . Find  $\hat{f}(s)$ .

**Solution:**

$$\hat{f}(s) = \int_0^{\infty} f(t)e^{-st}dt = \int_0^{\infty} e^{-st}dt = -\frac{1}{s}e^{-st} \Big|_0^{\infty} = \frac{1}{s} \quad (3)$$

The last equation is valid if

$$\int_0^{\infty} |f(t)e^{-\sigma t}|dt = \int_0^{\infty} |e^{-\sigma t}|dt < \infty \quad (4)$$

which means the real part of  $s$  must be positive.

Similarly if  $f(t) = e^{-\alpha t}$ ,  $t \geq 0$  then its Laplace transform is given by

$$\int_0^{\infty} f(t)e^{-st}dt = \int_0^{\infty} e^{-(\alpha+s)t}dt = -\frac{1}{s+\alpha}e^{-(s+\alpha)t} \Big|_0^{\infty} = \frac{1}{s+\alpha} \quad (5)$$

## Important Theorems of the Laplace Transform

**Theorem 1.** Let  $k$  be a constant and  $\hat{f}(s)$  be the Laplace transform of  $f(t)$ . Then

$$L[kf(t)] = k\hat{f}(s) \quad (6)$$

**Theorem 2.** Let  $\hat{f}_1(s)$  and  $\hat{f}_2(s)$  be the Laplace transforms of  $f_1(t)$  and  $f_2(t)$ , respectively. Then

$$L[f_1(t) \pm f_2(t)] = \hat{f}_1(s) \pm \hat{f}_2(s) \quad (7)$$

**Theorem 3.** Let  $\hat{f}(s)$  be the Laplace transforms of  $f(t)$  and  $f(0)$  is the limit of  $f(t)$  as  $t \rightarrow 0$ . Then

$$L\left[\frac{df(t)}{dt}\right] = s\hat{f}(s) - \lim_{t \rightarrow 0} f(t) = s\hat{f}(s) - f(0) \quad (8)$$

In general for higher derivatives of  $f(t)$

$$L\left[\frac{d^n f(t)}{dt^n}\right] = s^n \hat{f}(s) - s^{n-1} f(0) - s^{n-2} f^{(1)}(0) - \dots - f^{(n-1)}(0) \quad (9)$$

**Theorem 4.** Let  $\hat{f}(s)$  be the Laplace transforms of  $f(t)$  then

$$L \left[ \int_0^t f(\tau) d\tau \right] = \frac{\hat{f}(s)}{s} \quad (10)$$

**Theorem 5.** Shift in time

$$L [f(t - T)] = e^{-Ts} \hat{f}(s) \quad (11)$$

**Theorem 6.** Initial value theorem

$$\lim_{t \rightarrow 0} f(t) = \lim_{s \rightarrow \infty} s \hat{f}(s) \quad (12)$$

**Theorem 7.** Final value theorem

$$\lim_{t \rightarrow \infty} f(t) = \lim_{s \rightarrow 0} s \hat{f}(s) \quad (13)$$

**Theorem 8.** Let  $f_1(s)$  and  $\hat{f}_2(s)$  be the Laplace transforms of  $f_1(t)$  and  $f_2(t)$ , respectively, then

$$\hat{f}_1(s)\hat{f}_2(s) = L[f_1(t) * f_2(t)] = L\left[\int_0^t f_1(\tau)f_2(t-\tau)d\tau\right] \quad (14)$$

## Inverse Laplace Transform by Partial-Fraction Expansion

When Laplace transform solution is a rational function in  $s$ , it can be written as

$$\hat{f}(s) = \frac{Q(s)}{P(s)} \quad (15)$$

where  $P(s)$  and  $Q(s)$  are polynomials of  $s$ . It is assumed that the order of  $P(s)$  in  $s$  is greater than that of  $Q(s)$ .

### When All the Poles of $\hat{f}(s)$ Are Simple

In this case

$$\hat{f}(s) = \frac{Q(s)}{(s-s_1)(s-s_2)\cdots(s-s_n)} \quad (16)$$

where  $s_1 \neq s_2 \neq \cdots \neq s_n$ .

Applying the partial-fraction expansion, (16) can be written as

$$\hat{f}(s) = \frac{k_{s1}}{(s - s_1)} + \frac{k_{s2}}{(s - s_2)} + \cdots + \frac{k_{sn}}{(s - s_n)} \quad (17)$$

where  $k_{si}$  are given by

$$k_{si} = \left[ (s - s_i) \frac{Q(s)}{P(s)} \right] \Big|_{s=s_i} = \frac{Q(s_i)}{\prod_{j=1, j \neq i}^n (s_i - s_j)} \quad (18)$$

**Example 2:** Consider the function

$$\hat{f}(s) = \frac{5s + 3}{(s + 1)(s + 2)(s + 3)} \quad (19)$$

Find the partial-fraction expansion.

**Solution:** Here the partial-fraction expanded have the form

$$\hat{f}(s) = \frac{k_{-1}}{(s + 1)} + \frac{k_{-2}}{(s + 2)} + \frac{k_{-3}}{(s + 3)} \quad (20)$$

The coefficients  $k_{-1}$ ,  $k_{-2}$  and  $k_{-3}$  are determined by

$$k_{-1} = \left[ (s+1)\hat{f}(s) \right] \Big|_{s=-1} = \frac{5(-1)+3}{(2-1)(3-1)} = -1$$

$$k_{-2} = \left[ (s+2)\hat{f}(s) \right] \Big|_{s=-2} = \frac{5(-2)+3}{(1-2)(3-2)} = 7$$

$$k_{-3} = \left[ (s+3)\hat{f}(s) \right] \Big|_{s=-3} = \frac{5(-3)+3}{(1-3)(2-3)} = -6$$

Now by using the theorem 3 and example 1 we conclude that

$$f(t) = -e^{-t} + 7e^{-2t} - 6e^{-3t}, \quad t \geq 0 \quad (21)$$

### When Some Poles of $\hat{f}(s)$ Are of Multiple Order

If  $r$  of the  $n$  poles of  $\hat{f}(s)$  are identical, or, say, the pole at  $s = s_i$  is of multiplicity  $r$  we have

$$\hat{f}(s) = \frac{Q(s)}{(s-s_1)(s-s_2)\cdots(s-s_{n-r})(s-s_i)^r} \quad (22)$$



Then  $\hat{f}(s)$  can be expanded as follows

$$\hat{f}(s) = \underbrace{\frac{k_{s1}}{(s-s_1)} + \dots + \frac{k_{s(n-r)}}{(s-s_{n-r})}}_{(n-r) \text{ simple poles}} + \underbrace{\frac{a_1}{(s-s_i)} + \dots + \frac{a_r}{(s-s_i)^r}}_{r \text{ terms of repeated poles}} \quad (23)$$

$k_{s1}, \dots, k_{s(n-r)}$  can be evaluated using (18). For  $a_1, a_2, \dots, a_r$  we have

$$\begin{aligned} a_r &= \left[ (s-s_i)^r \hat{f}(s) \right] \Big|_{s=s_i} \\ a_{r-1} &= \frac{d}{ds} \left[ (s-s_i)^r \hat{f}(s) \right] \Big|_{s=s_i} \\ a_{r-2} &= \frac{1}{2!} \frac{d^2}{ds^2} \left[ (s-s_i)^r \hat{f}(s) \right] \Big|_{s=s_i} \\ &\vdots \\ a_1 &= \frac{1}{(r-1)!} \frac{d^{r-1}}{ds^{r-1}} \left[ (s-s_i)^r \hat{f}(s) \right] \Big|_{s=s_i} \end{aligned} \quad (24)$$

**Example 3:** Consider the function

$$\hat{f}(s) = \frac{1}{s(s+1)^3(s+2)} \quad (25)$$

Find the partial-fraction expansion.

**Solution:** Here the partial-fraction expanded form

$$\hat{f}(s) = \frac{k_0}{s} + \frac{k_{-2}}{(s+2)} + \frac{a_1}{(s+1)} + \frac{a_2}{(s+1)^2} + \frac{a_3}{(s+1)^3} \quad (26)$$

The coefficients of the simple poles are

$$k_0 = \left[ s\hat{f}(s) \right] \Big|_{s=0} = \frac{1}{2}$$
$$k_{-2} = \left[ (s+2)\hat{f}(s) \right] \Big|_{s=-2} = \frac{1}{2}$$

The coefficients corresponding to the third order pole are

$$\begin{aligned}a_3 &= \left[ (s+1)^3 \hat{f}(s) \right] \Big|_{s=-1} = -1 \\a_2 &= \frac{d}{ds} \left[ (s+2) \hat{f}(s) \right] \Big|_{s=-1} = \frac{d}{ds} \left[ \frac{1}{s(s+2)} \right] \Big|_{s=-1} = 0 \\a_1 &= \frac{1}{2!} \frac{d^2}{ds^2} \left[ (s+2) \hat{f}(s) \right] \Big|_{s=-1} = \frac{1}{2} \frac{d^2}{ds^2} \left[ \frac{1}{s(s+2)} \right] \Big|_{s=-1} = -1\end{aligned}$$

The complete partial-fraction expansion is

$$\hat{f}(s) = \frac{1}{2s} + \frac{1}{2(s+2)} - \frac{1}{s+1} - \frac{1}{(s+1)^3} \quad (27)$$

Now by using the theorem 3 and example 1 we conclude that

$$f(t) = \frac{1}{2} + \frac{1}{2}e^{-2t} - e^{-t} - \frac{1}{2}t^2e^{-t}, \quad t \geq 0 \quad (28)$$

Since  $L^{-1} [1/s^n] = t^{n-1}/(n-1)!$ .

## Z-Transform

Given a sequence  $f[k]$ ,  $k = 0, 1, \dots$  its Z-transform is given by

$$\hat{f}(z) = Z(f[k]) = \sum_{i=0}^{\infty} f[k]z^{-k} \quad (29)$$

where  $z$  is a complex quantity.

**Example 4:** Given  $f[k] = 1$ ,  $k = 0, 1, \dots$  find  $\hat{f}(z)$ .

**Solution:**

$$\hat{f}(z) = \sum_{i=0}^{\infty} f[k]z^{-k} = \sum_{i=0}^{\infty} z^{-k} = \frac{z}{1 - z}$$

for  $|z| > 1$ .

## Important Theorems of the Laplace Transform

**Theorem 1.** Let  $\hat{f}_1(z)$  and  $\hat{f}_2(z)$  be the Laplace transforms of  $f_1[k]$  and  $f_2[k]$ , respectively. Then

$$Z[f_1[k] \pm f_2[k]] = \hat{f}_1(z) \pm \hat{f}_2(z) \quad (30)$$

**Theorem 2.** If  $n > 0$  then

$$Z(f[k - n]) = z^{-n} \hat{f}(z) \quad (31)$$

and

$$Z(f[k + n]) = z^n \left( \hat{f}(z) - \sum_{i=1}^{n-1} f[i] z^{-i} \right) \quad (32)$$

**Theorem 3.** Let  $f_1(z)$  and  $\hat{f}_2(z)$  be the Laplace transforms of  $f_1[k]$  and  $f_2[k]$ , respectively, then

$$\hat{f}_1(z) \hat{f}_2(z) = Z(f_1[k] * f_2[k]) = Z\left(\sum_0^N f_1[k] f_2[k - N]\right) \quad (33)$$

**Inverse  $Z$ -transform** is obtained by first expanding  $\hat{f}(z)$  by partial fraction expansion of  $\hat{f}(Z)/z$  and then multiplying  $z$  both sides. This expansion gives recognizable terms, and  $Z$ -transform table is used to determine the corresponding  $f[k]$ .

## TIME FUNCTION

## LAPLACE TRANSFORM

Unit impulse  $\delta(t)$

Unit step  $U(t)$

$t$

$t^2$

$\frac{t^2}{2}$

$t^n$

$\frac{t^n}{n!}$

$e^{-\alpha t}$

$te^{-\alpha t}$

$1 - e^{-\alpha t}$

$\sin \omega t$

$e^{-\alpha t} \sin \omega t$

$\cos \omega t$

$e^{-\alpha t} \cos \omega t$

1

$\frac{1}{s}$

$\frac{1}{s^2}$

$\frac{1}{s^2}$

$\frac{1}{s^3}$

$\frac{1}{s^3}$

$\frac{1}{s^{n+1}}$

$\frac{1}{s^{n+1}}$

$\frac{1}{s+\alpha}$

$\frac{1}{s+\alpha}$

$\frac{1}{(s+\alpha)^2}$

$\frac{\alpha}{(s+\alpha)^2}$

$\frac{\alpha}{s(s+\alpha)}$

$\frac{\omega}{s^2+\omega^2}$

$\frac{\omega}{s^2+\omega^2}$

$\frac{\omega}{(s+\alpha)^2+\omega^2}$

$\frac{\omega}{(s+\alpha)^2+\omega^2}$

$\frac{s}{s^2+\omega^2}$

$\frac{s}{s^2+\omega^2}$

$\frac{s+\alpha}{(s+\alpha)^2+\omega^2}$

Table of Laplace transform

## TIME FUNCTION

Unit impulse  $\delta[k]$

Unit step  $U[k]$

$k$

$\frac{k^2}{2}$

$\frac{k^n}{n!}$

$\sin \omega k$

$\cos \omega k$

## Z-TRANSFORM

1

$\frac{z}{z-1}$

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

$\frac{z(z+1)}{2(z-1)^3}$

$\lim_{\alpha \rightarrow 0} \frac{(-1)^n}{n!} \frac{\partial^n}{\partial \alpha^n} \left( \frac{z}{z - e^{-\alpha}} \right)$

$\frac{z \sin w}{z^2 - 2z \cos w + 1}$

$\frac{z(z - \cos w)}{z^2 - 2z \cos w + 1}$

## Table of Z-transform

# Mathematical Descriptions of Systems

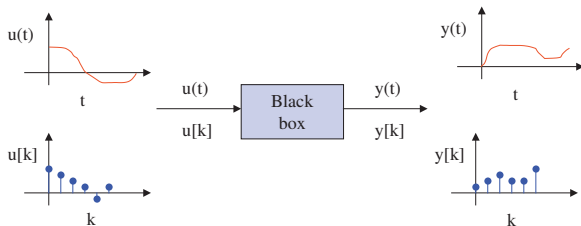


Fig. 5.1

## Some Abbreviations

**SISO** = Single Input Single Output

**MIMO** = Multiple Input Multiple Output

**SIMO** = Single Input Multiple Output



## Causality and Lumpedness

**Memoryless System:** If its output at  $t_0$  ( $\underline{y}(t_0)$ ) depends only on the inputs applied at  $t_0$ . Example: network of resistors.

**Causal System:** If its output at  $t_0$  ( $\underline{y}(t_0)$ ) depends on the inputs applied at  $t \leq t_0$ . Example: all physical systems.

**State of a System:** The state  $\underline{x}(t_0)$  of a system at  $t_0$  is the info. that, together with input  $\underline{u}(t)$ , for  $t \geq t_0$ , determines uniquely the output  $\underline{y}(t)$  for all  $t > t_0$ .

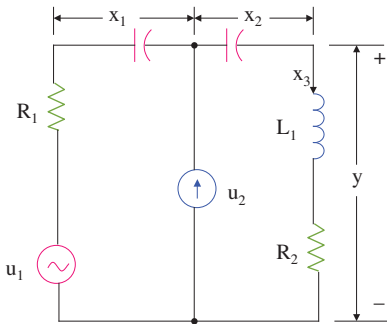


Fig. 5.2

**Lumped System:** If the number of state variables is finite. Otherwise the system is called **distributed system**.

**Linear System:** If for every  $t_0$  and for any two state-input-output pairs

$$\left. \begin{array}{l} \underline{x}_i(t_0) \\ \underline{u}_i(t), \quad t \geq t_0 \end{array} \right\} \rightarrow \underline{y}_i(t), \quad t \geq t_0, \quad i = 1, 2$$

we have the following

$$\left. \begin{array}{l} \alpha_1 \underline{x}_1(t_0) + \alpha_2 \underline{x}_2(t_0) \\ \alpha_1 \underline{u}_1(t) + \alpha_2 \underline{u}_2(t), \quad t \geq t_0 \end{array} \right\} \rightarrow \alpha_1 \underline{y}_1(t) + \alpha_2 \underline{y}_2(t), \quad t \geq t_0$$

where  $\alpha_1$  and  $\alpha_2$  are real constants. This is known as **superposition property**.

From the above we can also conclude that **response=zero-input response+ zero-state response**.

## Input-Output Description

We consider a SISO linear system. We consider a pulse  $\delta_{\Delta}(t - t_1)$  as shown in the Fig. 5.3 (a).

Then every input  $u(t)$  can be approximated as Fig. 5.3 (b), i.e.,

$$u(t) \approx \sum_i u(t_i) \delta_{\Delta}(t - t_i) \Delta$$

Let  $g_{\Delta}(t, t_i)$  be the output at time  $t$  excited by pulse  $\delta_{\Delta}(t - t_i)$  applied at  $t_i$ .

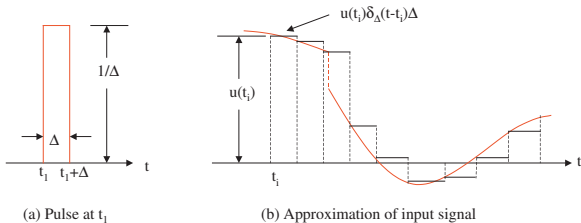


Fig. 5.3

Since the system is linear following input-output relations hold

$$\delta_{\Delta}(t - t_i) \rightarrow g_{\Delta}(t, t_i)$$

$$\delta_{\Delta}(t - t_i)u(t_i)\Delta \rightarrow g_{\Delta}(t, t_i)u(t_i)\Delta \quad (\text{homogeneity})$$

$$\sum_i \delta_{\Delta}(t - t_i)u(t_i)\Delta \rightarrow \sum_i g_{\Delta}(t, t_i)u(t_i)\Delta \quad (\text{additivity})$$

Thus the output  $y(t)$  can be approximated by

$$y(t) \approx \sum_i g_{\Delta}(t, t_i)u(t_i)\Delta \quad (34)$$

For  $\Delta \rightarrow 0$  the we have

$$y(t) = \int_{-\infty}^{+\infty} g(t, \tau)u(\tau)d\tau \quad (35)$$

Notice that  $g(t, \tau)$  is a function of two variables. Second variable denotes the time at which the impulse input was applied; the first one denotes the time at which the output is observed.

For a causal system  $g(t, \tau) = 0, \quad t < \tau$ .

The system is relaxed at  $t = t_0$  means its state at  $t = t_0$  is zero. Thus for a causal and relaxed at  $t_0$  system

$$y(t) = \int_{t_0}^t g(t, \tau) u(\tau) d\tau \quad (36)$$

If a linear system has  $p$  input terminals and  $q$  output terminals then

$$\underline{y}(t) = \int_{t_0}^t G(t, \tau) \underline{u}(\tau) d\tau \quad (37)$$

where

$$G(t, \tau) = \begin{bmatrix} g_{11}(t, \tau) & g_{12}(t, \tau) & \cdots & g_{1p}(t, \tau) \\ g_{21}(t, \tau) & g_{22}(t, \tau) & \cdots & g_{2p}(t, \tau) \\ \vdots & \vdots & \cdots & \vdots \\ g_{q1}(t, \tau) & g_{q2}(t, \tau) & \cdots & g_{qp}(t, \tau) \end{bmatrix}$$

## State-Space Description

A linear lumped system can be described by

$$\dot{\underline{x}}(t) = A(t)\underline{x}(t) + B(t)\underline{u}(t) \quad (38)$$

$$\underline{y}(t) = C(t)\underline{x}(t) + D(t)\underline{u}(t) \quad (39)$$

$$(40)$$

This is accepted as a fact!!!

## Linear Time-Invariant (LTI) Systems

A system is time invariant if for every state-input-output pair

$$\left. \begin{array}{l} \underline{x}(t_0) \\ \underline{u}(t), \quad t \geq t_0 \end{array} \right\} \rightarrow \underline{y}(t), \quad t \geq t_0$$

and any  $T$ , we have

$$\left. \begin{array}{l} \underline{x}(t_0 + T) \\ \underline{u}(t - T), \quad t \geq t_0 + T \end{array} \right\} \rightarrow \underline{y}(t - T), \quad t \geq t_0 + T \quad (41)$$

Example: Computers, calculators (gives the same result whenever they are used). What about human beings?

The opposite type of systems are called **time varying systems**.

Example: A burning rocket, its mass decreases with time. So does its the response to the force that drives it.

For a time invariant linear system the response of the system at  $t$  to an input applied at  $\tau$  is equivalent to

$$g(t, \tau) = g(t + T, \tau + T) = g(t - \tau, 0) = g(t - \tau)$$

In this case function  $g$  has only one variable which is the difference of the time at which the input was applied and the output was observed. The input output relation can be summarized by

$$y(t) = \int_0^t g(t - \tau)u(\tau)d\tau \quad (42)$$

Here the system is considered to be relaxed at time 0.

Using the fact that convolution in time domain gives product of functions after Laplace transform

$$y(t) = \int_0^t g(t - \tau)u(\tau)d\tau \quad \Rightarrow \quad \hat{y}(s) = \hat{g}(s)\hat{u}(s) \quad (43)$$

For a MIMO LTI system with  $p$  inputs and  $q$  outputs we have

$$\begin{bmatrix} \hat{y}_1(s) \\ \hat{y}_2(s) \\ \vdots \\ \hat{y}_q(s) \end{bmatrix} = \begin{bmatrix} \hat{g}_{11}(s) & \hat{g}_{12}(s) & \cdots & \hat{g}_{1p}(s) \\ \hat{g}_{21}(s) & \hat{g}_{22}(s) & \cdots & \hat{g}_{2p}(s) \\ \vdots & \vdots & \cdots & \vdots \\ \hat{g}_{q1}(s) & \hat{g}_{q2}(s) & \cdots & \hat{g}_{qp}(s) \end{bmatrix} \begin{bmatrix} \hat{u}_1(s) \\ \hat{u}_2(s) \\ \vdots \\ \hat{u}_p(s) \end{bmatrix}$$

Equivalent to  $\underline{\hat{y}}(s) = \hat{G}(s)\underline{\hat{u}}(s)$ .

Using Laplace transform the state-space equations can be rewritten as

$$\left. \begin{array}{l} \dot{\underline{x}}(t) = A\underline{x}(t) + B\underline{u}(t) \\ \underline{y}(t) = C\underline{x}(t) + D\underline{u}(t) \end{array} \right\} \Rightarrow \left\{ \begin{array}{l} s\underline{\hat{x}}(s) - \underline{x}(0) = A\underline{\hat{x}}(s) + B\underline{\hat{u}}(s) \\ \underline{\hat{y}}(s) = C\underline{\hat{x}}(s) + D\underline{\hat{u}}(s) \end{array} \right. \quad (44)$$



The previous equation implies

$$\begin{aligned}\hat{\underline{x}}(s) &= (sI - A)^{-1}\underline{x}(0) + (sI - A)^{-1}B\hat{\underline{u}}(s) \\ \hat{\underline{y}}(s) &= C(sI - A)^{-1}\underline{x}(0) + C(sI - A)^{-1}B\hat{\underline{u}}(s) + D\hat{\underline{u}}(s)\end{aligned}\quad (45)$$

### Example 5:

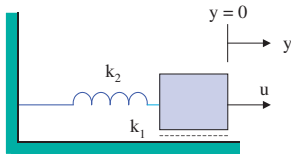


Fig. 5.4

Find out state-space relations and input output relation considering zero initial conditions.

**Solution:** Given  $k_1$  to be viscous friction coefficient,  $k_2$  to be the spring constant and mass of the block to be  $m$  we have

$$m\ddot{y} = u - k_1\dot{y} - k_2y \quad (46)$$

Applying Laplace transform we get

$$ms^2\hat{y}(s) = \hat{u}(s) - k_1s\hat{y}(s) - k_2\hat{y}(s)$$

which implies

$$\hat{y}(s) = \frac{1}{ms^2 + k_1s + k_2}\hat{u}(s)$$

If  $m = 1$ ,  $k_1 = 3$  and  $k_2 = 2$  then the impulse response

$$g(t) = L^{-1} \left[ \frac{1}{s^2 + 3s + 2} \right] = L^{-1} \left[ \frac{1}{s+1} - \frac{1}{s+2} \right] = e^{-t} - e^{-2t}$$

Let us select the displacement and velocity of the block as state variables, i.e.,  $x_1 = y$ ,  $x_2 = \dot{y}$ . We have

$$\dot{x}_1 = x_2 \quad \text{and} \quad m\dot{x}_2 = u - k_1x_2 - k_2x_1$$

They can be expressed as

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -k_2/m & -k_1/m \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 1/m \end{bmatrix} u(t)$$

The output is related with the state variables as follows

$$y(t) = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}$$

### Example 6:

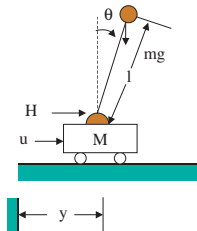


Fig. 5.5

Find out state-space relations and input output relation considering zero initial conditions.

**Solution:** From Newton's law of linear movement we have

$$M\ddot{y} = u - H = u - m\frac{d^2}{dt^2}(y + l\cos\theta) = m\ddot{y} + ml\ddot{\theta}\cos\theta - ml(\dot{\theta})^2\sin\theta$$

The application of Newton's law to the rotational movement gives

$$mgl \sin \theta = ml^2 \ddot{\theta} + m\ddot{y}l \cos \theta$$

For small  $\theta$  and  $\dot{\theta}$  we have  $\sin \theta = \theta$ ,  $\cos \theta = 1$  and terms having  $\theta^2$ ,  $(\dot{\theta})^2$ ,  $\theta\dot{\theta}$  and  $\theta\ddot{\theta}$  can be neglected. We have

$$M\ddot{y} = u - mg\theta$$

$$Ml\ddot{\theta} = (M + m)g\theta - u$$

For zero initial condition Laplace transform gives

$$Ms^2\hat{y}(s) = \hat{u}(s) - mg\hat{\theta}(s)$$

$$Mls^2\hat{\theta}(s) = (M + m)g\hat{\theta}(s) - \hat{u}(s)$$

The transfer functions  $\hat{g}_{yu}(s)$  and  $\hat{g}_{\theta u}(s)$  can be obtained from the above relations.

Let the state variables be  $x_1 = y$ ,  $x_2 = \dot{y}$ ,  $x_3 = \theta$ ,  $x_4 = \dot{\theta}$ . Then from this relations we have

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \\ \dot{x}_3(t) \\ \dot{x}_4(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & -mg/M & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & (M+m)g/Ml & 0 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \\ x_4(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 1/M \\ 0 \\ -1/Ml \end{bmatrix} u(t)$$

and  $y = [1 \ 0 \ 0 \ 0]\underline{x}$ .

### Example 7:

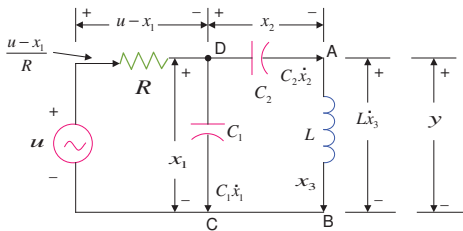


Fig. 5.6

Find out state-space relations.

**Solution:** We consider voltages across the capacitors and currents through the inductors as state variables.

Equating currents at A and D and voltage in the loop ABCD we get

$$\frac{u - x_1}{R} = C_1 \dot{x}_1 + C_2 \dot{x}_2 = C_1 \dot{x}_1 + x_3$$

$$C_2 \dot{x}_2 = x_3$$

$$L \dot{x}_3 = x_1 - x_2$$

And the output is given by

$$y = L \dot{x}_3 = x_1 - x_2 \quad (47)$$

Thus

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \\ \dot{x}_3(t) \end{bmatrix} = \begin{bmatrix} -1/RC_1 & 0 & -1/C_1 \\ 0 & 0 & 1/C_2 \\ 1/L & -1/L & 0 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{bmatrix} + \begin{bmatrix} 1/RC_1 \\ 0 \\ 0 \end{bmatrix} u(t)$$

and  $y = [1 \ -1 \ 0] \underline{x} + 0.u.$

## Example 7:

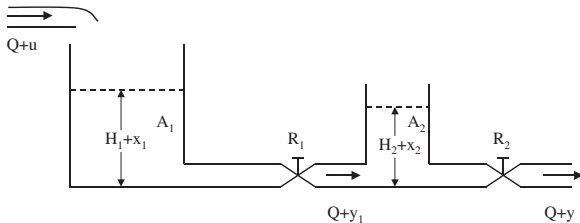


Fig. 5.7

It is assumed that under normal operation, inflows and outflows of both tanks are equal to  $Q$  and levels are  $H_1$  and  $H_2$ .

Let  $u$  be the inflow perturbation of the first tank which causes a change of liquid level  $x_1$  and outflow  $y_1$  for this tank, and level variation  $x_2$  and flow change  $y_2$  for the second tank.

It is assumed that  $y_1 = \frac{x_1 - x_2}{R_1}$  and  $y = x_2 R_2$ .

Find out state-space relations.

**Solution:** Changes in the liquid levels are governed by

$$A_1 dx_1 = (u - y_1)dt \quad \text{and} \quad A_2 dx_2 = (y_1 - y)dt$$

Thus

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{bmatrix} = \begin{bmatrix} -1/A_2 R_2 & 1/A_1 R_1 \\ 1/A_2 R_1 & -(1/A_2 R_1 + 1/A_2 R_2) \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \begin{bmatrix} 1/A_1 \\ 0 \end{bmatrix} u(t)$$

and  $y = [0 \ 1/R_2]\underline{x} + 0.u$ .

## Discrete Time Systems

For a linear discrete time system the output sequence  $y[k]$  is input sequence  $u[k]$  and the system response  $g[k, m]$  are related as

$$y[k] = \sum_{m=-\infty}^{+\infty} g[k, m]u[m] \quad (48)$$



For a causal system  $g[k, m] = 0, \quad k < m$ . In addition if the system is relaxed at  $k_0$ , we have

$$y[k] = \sum_{m=k_0}^k g[k, m]u[m] \quad (49)$$

For a time invariant system we can write

$$y[k] = \sum_{m=0}^k g[k - m]u[m] \quad (50)$$

Z-transform of the above equation gives

$$\hat{y}(z) = \hat{g}(z)\hat{u}(z) \quad (51)$$

For every lumped discrete time system we have

$$\begin{aligned} \underline{x}[k + 1] &= A[k]\underline{x}[k] + B[k]\underline{u}[k] \\ y[k] &= C[k]\underline{x}[k] + D[k]\underline{u}[k] \end{aligned} \quad (52)$$