Lecture 12

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State-Space Representation of DT Systems

State-space representation of a (causal & finite dimensional) LTI CT system is given by

Let
$$x(t) \in \mathbb{R}^n$$
, $y(t) \in \mathbb{R}^m$, $u(t) \in \mathbb{R}^r$,
$$\dot{x}(t) = Ax(t) + Bu(t),$$
$$y(t) = Cx(t) + Du(t),$$
where $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times r}$, $C \in \mathbb{R}^{m \times n}$, $D \in \mathbb{R}^{m \times r}$

State-space representation of a (causal & finite dimensional) LTI DT system is given by

Let
$$x[k] \in \mathbb{R}^n$$
, $y[k] \in \mathbb{R}^m$, $u[k] \in \mathbb{R}^r$,
$$x[k+1] = Gx[k] + Hu[k],$$

$$y[k] = Cx[k] + Du[k],$$
 where $G \in \mathbb{R}^{n \times n}$, $H \in \mathbb{R}^{n \times r}$, $C \in \mathbb{R}^{m \times n}$, $D \in \mathbb{R}^{m \times r}$

Depending on the values of m and r we have

- m=r=1, the system represents a SISO system
- ullet m>1 , r<1, the system represents a MIMO system
- m=1, r>1, the system represents a MISO system
- m > 1 , r = 1, the system represents a SIMO system

for both CT and DT cases.

State property of CT state-space models: Given the initial time, t_0 and state $x(t_0)$ and input u(t) for $t_0 \ge t < t_f$ (with t_0 & t_f arbitrary), we can compute the output y(t) for $t_0 \ge t < t_f$ and the state x(t) for $t_0 < t \ge t_f$.

State property of DT state-space models: Given the state vector x[k] and input u[k] at an arbitrary time k, we can compute the present output, y[k], and next state x[k+1].

Note that both definitions are not limited to LTI state-space models. Nonlinear and time-varying state-space models also are based on this definition.

When a state-space representation includes minimum number of state variables, the representation is called minimal.

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Canonical State-Space Realizations of (SISO) DT Systems

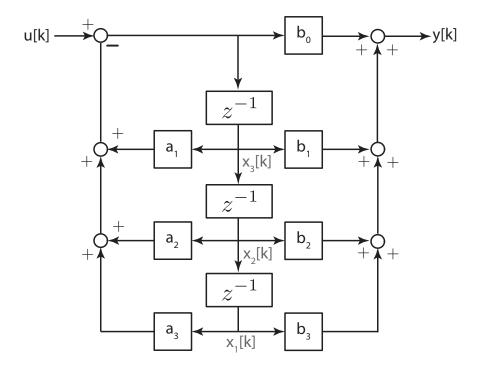
Reachable/Controllable Canonical Form

For the sake of clarity let's assume that the system that we would like to represent is a third order DT system with the following difference equation and transfer function

$$y[k] = -a_1 y[k-1] - a_2 y[k-2] - a_3 y[k-3] + b_0 x[k] + b_1 x[k-1] + b_2 x[k-2] + b_3 x[k-3]$$

$$Y(z) = \frac{b_0 + b_1 z^{-1} + b_2 z^{-2} + b_3 z^{-3}}{1 + a_1 z^{-1} + a_2 z^{-2} + a_3 z^{-3}} X(z)$$

We know that following block diagram realizes this system structure with minimum number of delay elements and it is a canonical realization. Delay operation is directly related with state and state evolution concept.



If we label the signals as given in the Figure, state evolution equations can be derived as

$$X_1(z) = X_2(z)z^{-1} \rightarrow x_1[k+1] = x_2[k]$$

 $X_2(z) = X_3(z)z^{-1} \rightarrow x_2[k+1] = x_3[k]$

$$X_2(z) = X_3(z)z^{-1} \rightarrow x_2[k+1] = x_3[k]$$

$$X_3(z) = (U(z) - (X_1(z)a_3 + X_2(z)a_2 + X_3(z)a_1))z^{-1} \rightarrow x_3[k+1] = u[k] + x_1[k](-a_3) + x_2[k](-a_2) + x_3[k](-a_1)$$

where as output equation can be derived as

$$y[k] = b_1 x_3[k] + b_2 x_2[k] + b_3 x_1[k] b_0 u[k] - b_0 (a_1 x_3[k] + a_2 x_2[k] + a_3 x_1[k])$$

= $b_0 u[k] + (b_3 - b_0 a_3) x_1[k] + (b_2 - b_0 a_2) x_2[k] + (b_1 - b_0 a_1) x_3[k]$

If we gather these equations, we can obtain the state space form

$$\mathbf{x}[k+1] = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -a_3 & -a_2 & -a_1 \end{bmatrix} \mathbf{x}[k] + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} u[k]$$
$$y[k] = \begin{bmatrix} (b_3 - b_0 a_3) & (b_2 - b_0 a_2) & (b_1 - b_0 a_1) \end{bmatrix} + b_0 u[k]$$

where

$$\mathbf{x} = \begin{bmatrix} x_1[k] \\ x_2[k] \\ x_1[k] \end{bmatrix} , A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -a_3 & -a_2 & -a_1 \end{bmatrix} , B = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$
$$C = \begin{bmatrix} (b_3 - b_0 a_3) & (b_2 - b_0 a_2) & (b_1 - b_0 a_1) \end{bmatrix} , D = b_0$$

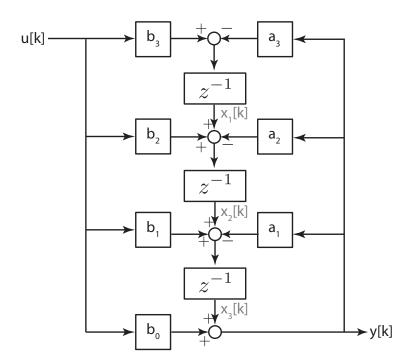
The form obtained with this approach is called reachable/controllable canonical form.

For a general n^{th} order system reachable/controllable canonical form has the following A , B , C , & D matrices

$$A = \begin{bmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & & \vdots \\ 0 & 0 & 0 & \cdots & 1 \\ -a_n & -a_{n-1} & -a_{n-2} & \cdots & -a_1 \end{bmatrix}, B = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix}$$
$$C = \begin{bmatrix} (b_n - b_0 a_n) & (b_{n-1} - b_0 a_{n-1}) & \cdots & (b_2 - b_0 a_2) & (b_1 - b_0 a_1) \end{bmatrix}, D = b_0$$

Observable Canonical Form

We also learnt a different type of canonical minimal realization which is illustrated in the Figure below



If we label the signals as given in the Figure, state evolution equations can be derived as

$$X_{1}(z) = (b_{3}U(z) - a_{3}(U(z)b_{0} + X_{3}(z)))z^{-1} \rightarrow x_{1}[k+1] = b_{3}u[k] - a_{3}(u[k]b_{0} + x_{3}[k])$$

$$X_{2}(z) = (b_{2}U(z) + X_{1}(z) - a_{2}(U(z)b_{0} + X_{3}(z)))z^{-1} \rightarrow x_{2}[k+1] = b_{2}u[k] + x_{1}[k] - a_{2}(u[k]b_{0} + x_{3}[k])$$

$$X_{3}(z) = (b_{1}U(z) + X_{2}(z) - a_{1}(U(z)b_{0} + X_{3}(z)))z^{-1} \rightarrow x_{3}[k+1] = b_{1}u[k] + x_{2}[k] - a_{1}(u[k]b_{0} + x_{3}[k])$$

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where as output equation can be derived as

$$y[k] = b_0 u[k] + x_3[k]$$

If we combine the state and output equations, we can obtain the state space form as

$$\mathbf{x}[k+1] = \begin{bmatrix} 0 & 0 & -a_3 \\ 1 & 0 & -a_2 \\ 0 & 1 & -a_1 \end{bmatrix} \mathbf{x}[k] + \begin{bmatrix} b_3 - b_0 a_3 \\ b_2 - b_0 a_2 \\ b_1 - b_0 a_1 \end{bmatrix} u[k]$$
$$y[k] = \begin{bmatrix} 0 & 0 & 1 \end{bmatrix} + b_0 u[k]$$

where

$$\mathbf{x} = \left[\begin{array}{c} x_1[k] \\ x_2[k] \\ x_1[k] \end{array} \right] \quad , \quad A = \left[\begin{array}{ccc} 0 & 0 & -a_3 \\ 1 & 0 & -a_2 \\ 0 & 1 & -a_1 \end{array} \right] \quad , \quad B = \left[\begin{array}{c} b_3 - b_0 a_3 \\ b_2 - b_0 a_2 \\ b_1 - b_0 a_1 \end{array} \right] \quad , \quad C = \left[\begin{array}{ccc} 0 & 0 & 1 \end{array} \right] \quad , \quad D = b_0$$

The form obtained with this approach is called observable canonical form.

For a general n^{th} order system observable canonical form has the following A, B, C, & D matrices

$$A = \begin{bmatrix} 0 & 0 & \cdots & 0 & -a_n \\ 1 & 0 & \cdots & 0 & -a_{n-1} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 & -a_2 \\ 0 & 0 & \cdots & 1 & -a_1 \end{bmatrix} , \quad B = \begin{bmatrix} (b_n - b_0 a_n) \\ (b_{n-1} - b_0 a_{n-1}) \\ \vdots \\ (b_2 - b_0 a_2) \\ (b_1 - b_0 a_1) \end{bmatrix}$$

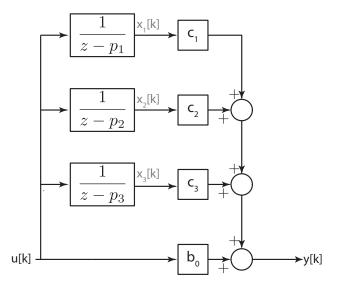
$$C = \begin{bmatrix} 0 & 0 & \cdots & 0 & 1 \end{bmatrix} , \quad D = b_0$$

Diagonal Canonical Form

If the pulse transfer function of the system has distinct poles, we can expand it using partial fraction expansion

$$Y(z) = \frac{b_0 + b_1 z^{-1} + b_2 z^{-2} + b_3 z^{-3}}{1 + a_1 z^{-1} + a_2 z^{-2} + a_3 z^{-3}} X(z)$$
$$= b_0 + \frac{c_1}{z - p_1} + \frac{c_2}{z - p_2} \frac{c_3}{z - p_3}$$

It is possible to realize this expanded form in block diagram form as given in the Figure below.



Now let's concentrate on the candidate "state variables" and try to write state evaluation equations

$$X_1(z) = \frac{1}{z - p_1} U(z) \quad \to \quad x_1[k+1] = p_1 x_1[k] + u[k]$$

$$X_2(z) = \frac{1}{z - p_2} U(z) \quad \to \quad x_2[k+1] = p_2 x_2[k] + u[k]$$

$$X_3(z) = \frac{1}{z - p_3} U(z) \quad \to \quad x_3[k+1] = p_3 x_3[k] + u[k]$$

where as output equation can be derived as

$$y[k] = b_0 u[k] + c_1 x_1[k] + c_2 x_2[k] + c_3 x_3[k]$$

If we combine the state and output equations, we can obtain the state space form as

$$\mathbf{x}[k+1] = \begin{bmatrix} p_1 & 0 & 0 \\ 0 & p_2 & 0 \\ 0 & 0 & p_3 \end{bmatrix} \mathbf{x}[k] + \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} u[k]$$
$$y[k] = \begin{bmatrix} c_1 & c_2 & c_3 \end{bmatrix} + b_0 u[k]$$

where

$$\mathbf{x} = \left[\begin{array}{c} x_1[k] \\ x_2[k] \\ x_1[k] \end{array} \right] \quad , \quad A = \left[\begin{array}{ccc} p_1 & 0 & 0 \\ 0 & p_2 & 0 \\ 0 & 0 & p_3 \end{array} \right] \quad , \quad B = \left[\begin{array}{c} 1 \\ 1 \\ 1 \end{array} \right] \quad , \quad C = \left[\begin{array}{ccc} c_1 & c_2 & c_3 \end{array} \right] \quad , \quad D = b_0$$

The form obtained with this approach is called diagonal canonical form. Obviously, this form is not aplicable for systems that has repeated roots.

For a general n^{th} order system with distinct roots diagonal canonical form has the following A , B , C , & D

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matrices

$$A = \begin{bmatrix} p_1 & 0 & \cdots & 0 & 0 \\ 0 & p_2 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & p_{n-1} & 0 \\ 0 & 0 & \cdots & 0 & p_n \end{bmatrix} , B = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \\ 1 \end{bmatrix}$$
$$C = \begin{bmatrix} c_1 & c_2 & \cdots & c_{n-1} & c_n \end{bmatrix} , D = b_0$$

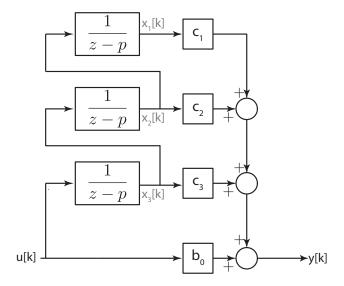
Jordan Canonical Form

Generalization of diagonal canonical farm is called Jordan canonical form which handles repeated roots.

In Jordan form the distinct roots has the same structure with Diagonal canonical form. Let's assume that the 3^{rd} order pulse transfer function has three repeated roots. In this case, we can expand it using partial fraction expansion

$$Y(z) = \frac{b_0 + b_1 z^{-1} + b_2 z^{-2} + b_3 z^{-3}}{1 + a_1 z^{-1} + a_2 z^{-2} + a_3 z^{-3}} X(z)$$
$$= b_0 + \frac{c_1}{(z - p)^3} + \frac{c_2}{(z - p)^2} \frac{c_3}{z - p}$$

It is possible to realize this expanded form in block diagram form as given in the Figure below.



Now let's concentrate on the candidate "state variables" and try to write state evaluation equations

$$X_1(z) = \frac{1}{z - p} X_2(z) \rightarrow x_1[k + 1] = p \ x_1[k] + x_2[k]$$

$$X_2(z) = \frac{1}{z - p} X_3(z) \rightarrow x_2[k + 1] = p \ x_2[k] x_3[k]$$

$$X_3(z) = \frac{1}{z - p} U(z) \rightarrow x_3[k + 1] = p \ x_3[k] + u[k]$$

where as output equation can be derived as

$$y[k] = b_0 u[k] + c_1 x_1[k] + c_2 x_2[k] + c_3 x_3[k]$$

If we combine the state and output equations, we can obtain the state space form as

$$\mathbf{x}[k+1] = \begin{bmatrix} p & 1 & 0 \\ 0 & p & 1 \\ 0 & 0 & p \end{bmatrix} \mathbf{x}[k] + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} u[k]$$
$$y[k] = \begin{bmatrix} c_1 & c_2 & c_3 \end{bmatrix} + b_0 u[k]$$

where

$$\mathbf{x} = \begin{bmatrix} x_1[k] \\ x_2[k] \\ x_1[k] \end{bmatrix} \quad , \quad A = \begin{bmatrix} p & 1 & 0 \\ 0 & p & 1 \\ 0 & 0 & p \end{bmatrix} \quad , \quad B = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \quad , \quad C = \begin{bmatrix} c_1 & c_2 & c_3 \end{bmatrix} \quad , \quad D = b_0$$

A, B, & C forms a Jordan block.

For a general n^{th} order system a Jordan block with m repeated roots inside a stat-space representation in Jordan canonical form looks like

$$A = \begin{bmatrix} \frac{\ddots & & & & & \\ \bar{p} & 1 & \cdots & 0 & 0 & \\ 0 & \bar{p} & \cdots & 0 & 0 & \\ & & \ddots & & & \\ 0 & 0 & \cdots & \bar{p} & 1 & \\ 0 & 0 & \cdots & 0 & \bar{p} & \\ \hline & & & \ddots & & \\ \end{bmatrix} , B = \begin{bmatrix} \frac{\vdots}{0} \\ 0 \\ 0 \\ \vdots \\ 0 \\ 1 \\ \vdots \end{bmatrix}$$

$$C = \begin{bmatrix} \cdots & c_1 & c_2 & \cdots & c_{n-1} & c_n & \cdots \end{bmatrix}$$

Similarity Transformations

Consider the state-space representation of a given DT system

$$x[k+1] = Gx[k] + Hu[k]$$
$$y[k] = Cx[k] + Du[k]$$

Let's define a new "state-vector" \hat{x} such that

$$x[k] = P\hat{x}[k]$$
 where $P \in \mathbb{R}^{n \times n} \det(P) \neq 0$

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Then we can transform the state-space equations using P as

$$\begin{split} P^{-1}\hat{x}[k+1] &= GP^{-1}\hat{x}[k] + Hu[k] \quad , \quad y[k] &= CP^{-1}x[k] + Du[k] \\ \hat{x}[k+1] &= PGP^{-1}\hat{x}[k] + PHu[k] \quad , \quad y[k] &= CP^{-1}x[k] + Du[k] \end{split}$$

The "new" state-space representation is obtained as

$$\begin{split} \hat{x}[k+1] &= \hat{G}\hat{x}[k] + \hat{H}u[k] \\ y[k] &= \hat{C}x[k] + \hat{D}u[k] \\ \hat{G} &= PGP^{-1} \ , \ \hat{H} = PH \ , \ \hat{C} = CP^{-1} \ , \ \hat{D} = D \end{split}$$

Since there exist infinitely many non-singular $n \times n$ matrices, for a given LTI DT systems, there exist infinitely many different but equivalent state-space representations.

Example: Show that $A \in \mathbb{R}^{n \times n}$ and $P^{-1}AP$, where $P\mathbb{R}^{n \times n}$ and $\det(P) \neq = 0$, have the same characteristic equation

Solution:

$$\det (\lambda I - P^{-1}AP) = \det (\lambda P^{-1}IP - P^{-1}AP)$$

$$= \det (P^{-1}(\lambda I - A)P)$$

$$= \det (P^{-1}) \det (\lambda I - A) \det (P)$$

$$= \det (P^{-1}) \det (P) \det (\lambda I - A)$$

$$\det (\lambda I - P^{-1}AP) = \det (\lambda I - A)$$

Obtaining Transfer Functions from a State-Space Representation

Let's consider the following general state-space form

$$x[k+1] = Gx[k] + Hu[k]$$
$$y[k] = Cx[k] + Du[k]$$

In order to obtain transfer function form, we assume that initial conditions are zero. Under this assumption, let's take z-transform of both equations

$$\begin{split} zX(z) &= GX(z) + HU(z) \quad , \quad Y(z) = CX(z) + DU(z) \\ (zI - G) \, X(z) &= HU(z) \\ X(z) &= (zI - G)^{-1} \, HU(z) \\ Y(z) &= C \, (zI - G)^{-1} \, HU(z) + DU(z) \\ Y(z) &= \left[C \, (zI - G)^{-1} \, H + D \right] U(z) \\ T(z) &= \left[C \, (zI - G)^{-1} \, H + D \right] \end{split}$$

If the system is a SISO system, then T(z) is a transfer function, where as for MIMO case T(z) becomes a transfer function matrix. Note that $(zI - G)^{-1}$ is invertible for all $z \in \mathbb{C}$ except the eigenvalues of G.

Example: Let p be a pole of T(z), show that p is also an eigenvalue of G.

Solution: Let

$$T(z) = \frac{n(z)}{d(z)}$$

If p is a pole of T(z), then $d(z)|_p = 0$. Now let's analyze the dependence of T(z) to the state-space form.

$$T(z) = \left[C (zI - G)^{-1} H + D\right]$$
$$(zI - G)^{-1} = \frac{\operatorname{Adj} (zI - G)}{\det (zI - G)}$$
$$T(z) = \frac{C\operatorname{Adj} (zI - G) H + D\det (zI - G)}{\det (zI - G)}$$

If p is a pole of T(z), then

$$\det\left(zI - G\right)|_{z=p} = 0$$

Obviously p is an eigenvalue of G.

Invariance of Transfer Functions Under Similarity Transformation

Consider the two different state-space representations

$$x[k+1] = Gx[k] + Hu[k]$$
 $\hat{x}[k+1] = \hat{G}\hat{x}[k] + \hat{H}u[k]$
 $y[k] = Cx[k] + Du[k]$ $y[k] = \hat{C}x[k] + \hat{D}u[k]$

where they are related with the following similarity transformation

$$x[k] = P\hat{x}[k] \ , \ \hat{G} = PGP^{-1} \ , \ \hat{H} = PH \ , \ \hat{C} = CP^{-1} \ , \ \hat{D} = D$$

Let's compute the transfer function for the second representation

$$\hat{T}(z) = \left[\hat{C}\left(zI - \hat{G}\right)^{-1}\hat{H} + \hat{D}\right]$$

$$= \left[CP^{-1}\left(zI - PGP^{-1}\right)^{-1}PH + D\right]$$

$$= \left[CP^{-1}\left(P\left(zI - G\right)P^{-1}\right)^{-1}PH + D\right]$$

$$= \left[CP^{-1}P\left(zI - G\right)^{-1}P^{-1}PH + D\right]$$

$$= \left[C\left(zI - G\right)^{-1}H + D\right]$$

$$\hat{T}(z) = T(z)$$

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Solution of Discrete-Time State-Space Equations

Let's first assume that u[k] = 0, and find un-driven (homogeneous) response.

$$x[k+1] = Gx[k]$$
$$y[k] = Cx[k]$$

Unlike CT systems we can compute the response iteratively

$$x[1] = Gx[0]$$

$$x[2] = Gx[1] = G^{2}x[0]$$

$$x[3] = Gx[2] = G^{3}x[0]$$

$$\vdots$$

$$x[k] = Gx[k-1] = G^{k}x[0] \quad , \quad y[k] = CG^{k}x[0]$$

It is easy to see that for $k, p \in \mathbb{Z}$ where k > p

$$x[k] = G^{k-p}x[p]$$

Let $\Psi(k) = G^k$, then this matrix of functions solves the homogeneous difference equation

$$x[k] = Gx[k]$$

$$x[k] = \Psi[k[x[0]$$

$$x[k] = \Psi[k-p]x[p]$$

$$x[k+m] = \Psi[k+m-m]x[m] = \Psi[k]$$

Let $\Psi[k]$ is called the state-transition matrix. Now let's consider input-only state response (i.e. x[0] = 0).

$$\begin{split} x[k+1] &= Gx[k] + Hu[k] \\ x[1] &= Hu[0] \\ x[2] &= Gx[1] + Hu[1] = GHu[0] + Hu[1] \\ x[3] &= Gx[2] + Hu[2] = G^2Hu[0] + GHu[1] + Hu[2] \\ x[4] &= Gx[3] + Hu[3] = G^3Hu[0] + G^2Hu[1] + GHu[2] + Hu[3] \\ &\vdots \\ x[k] &= Gx[k-1] + Hu[k-1] \\ &= G^{k-1}Hu[0] + G^{k-2}Hu[1] + \dots + GHu[k-2] + Hu[k-1] \\ &= \left[G^{k-1}H \mid G^{k-2}H \mid \dots \mid GH \mid H \mid \right] \begin{bmatrix} u[0] \\ u[1] \\ \vdots \\ u[k-2] \\ u[k-1] \end{bmatrix} \\ &= \sum_{j=0}^{k-1} G^{k-j-1}Hu[j] \\ &= \sum_{j=0}^{k-1} G^jHu[k-j-1] \end{split}$$

Given that $\Psi[k] = G^k$

$$x[k] = \sum_{j=0}^{k-1} \Psi[k-j-1] Hu[j]$$
$$= \sum_{j=0}^{k-1} \Psi[j] Hu[k-j-1]$$

If we combine homegeneous and driven responses we can simply obtain

$$x[k] = \Psi[k]x[0] + \sum_{j=0}^{k-1} \Psi[k-j-1]Hu[j]$$
$$= \Psi[k]x[0] + \sum_{j=0}^{k-1} \Psi[j]Hu[k-j-1]$$

whereas output at time k has the form

$$y[k] = C\Psi[k]x[0] + C\left(\sum_{j=0}^{k-1} \Psi[k-j-1]Hu[j]\right) Du[k]$$
$$= C\Psi[k]x[0] + C\left(\sum_{j=0}^{k-1} \Psi[j]Hu[k-j-1]\right) Du[k]$$

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Z-domain Solution of State-Space Equations

We already computed the transfer function under zero initial conditions.

$$Y(z) = \left[C \left(zI - G \right)^{-1} H + D \right] U(z)$$

Now let's compute the response to initial condition in Z-domain.

$$\mathcal{Z}[x[k+1]] = \mathcal{Z}[Gx[k]]$$

$$zX(z) - zx[0] = GX(z)$$

$$(zI - G)X(z) = zX(z)$$

$$X(z) = z(zI - G)^{-1}x[0]$$

Similarly Y(z) takes the form

$$Y(z) = zC (zI - G)^{-1} x[0]$$

We can also observe that

$$\mathcal{Z}[\Psi[k]] = \mathcal{Z}[G^k] = z (zI - G)^{-1}$$

 $\mathcal{Z}^{-1}[z (zI - G)^{-1}] = \Psi[k] = G^k$

If we expand $z(zI-G)^{-1}$ by long "division" we can also observe the relation in z-domain and time doman expressions from a different perspective

$$z (zI - G)^{-1} = I + z^{-1}G + z^{-2}G^2 + z^{-3}G^3 + \cdots$$

 $\mathcal{Z} \left[z (zI - G)^{-1} \right] = \{ I, G, G^2, G^3, \cdots \}$

Example: Consider the following state-space representation

$$x[k+1] = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1/2 & 0 \\ 0 & 0 & -1 \end{bmatrix} x[k] + \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} u[k]$$
$$y[k] = \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} x[k]$$

• Compute the closed form expression $\Psi[k]$ using the time expression Solution: The state-space representation is in Diagonal canonical form

$$\Psi[k] = G^k = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1/2 & 0 \\ 0 & 0 & -1 \end{bmatrix}^k = \begin{bmatrix} 1^k & 0 & 0 \\ 0 & (1/2)^k & 0 \\ 0 & 0 & (-1)^k \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & (1/2)^k & 0 \\ 0 & 0 & (-1)^k \end{bmatrix}$$

• Compute the closed form expression $\Psi[k]$ using the z-domain solution method Solution:

$$\begin{split} \Psi[k] &= \mathcal{Z}^{-1} \left[z \left(zI - G \right)^{-1} \right] \\ &= \mathcal{Z}^{-1} \left[z \left(\begin{bmatrix} z - 1 & 0 & 0 \\ 0 & (z - 1/2) & 0 \\ 0 & 0 & z + 1 \end{bmatrix} \right)^{-1} \right] \\ &= \mathcal{Z}^{-1} \left[\begin{bmatrix} \frac{z}{z - 1} & 0 & 0 \\ 0 & \frac{z}{z - 1/2} & 0 \\ 0 & 0 & \frac{z}{z + 1} \end{bmatrix} \right] \\ &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & (1/2)^k & 0 \\ 0 & 0 & (-1)^k \end{bmatrix} \quad \text{for } k \ge 0 \end{split}$$

• Compute the impulse response of the system from the time domain solution Solution:

$$\begin{split} x[k] &= G^{k-1} H u[0] \quad \text{for } k > 0 \\ y[k] &= C G^{k-1} H \quad \text{for } k > 0 \\ &= \left[\begin{array}{ccc} 1 & 2 & 3 \end{array} \right] \left[\begin{array}{ccc} 1 & 0 & 0 \\ 0 & (1/2)^{k-1} & 0 \\ 0 & 0 & (-1)^{k-1} \end{array} \right] \left[\begin{array}{c} 1 \\ 1 \\ 1 \end{array} \right] \\ &= \left[\begin{array}{ccc} 1 & 2 & 3 \end{array} \right] \left[\begin{array}{ccc} 1 \\ (1/2)^{k-1} \\ (-1)^{k-1} \end{array} \right] \\ y[k] &= 1 + 2(1/2)^{k-1} + 3(-1)^{k-1} \quad \text{for } k > 0 \end{split}$$

• Compute the transfer function $\frac{Y(z)}{U(z)}$

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Solution:

$$T(z) = C (zI - G)^{-1} H$$

$$= \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} z - 1 & 0 & 0 \\ 0 & z - 1/2 & 0 \\ 0 & 0 & z + 1 \end{bmatrix}^{-1} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} \frac{1}{z-1} & 0 & 0 \\ 0 & \frac{1}{z-1/2} & 0 \\ 0 & 0 & \frac{1}{z+1} \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} \frac{1}{z-1} \\ \frac{1}{z-1/2} \\ \frac{1}{z+1} \end{bmatrix}$$

$$T(z) = \frac{1}{z-1} + \frac{2}{z-1/2} + \frac{3}{z+1}$$

• Compute the inverse Z-transform of the transfer function Solution:

$$t[k] = \mathcal{Z}^{-1} \left[\frac{1}{z-1} + \frac{2}{z-1/2} + \frac{3}{z+1} \right]$$

= $\left(1 + 2(1/2)^{k-1} + 3(-1)^{k-1} \right) h[k-1]$

where h[k] is the unit step function