

## "State-Space Representation"

\* Review to some elementary Matrices notation and operations:-

- Matrix: is a set of elements arranged in Rows and Columns.

Ex:  $A = \begin{pmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{m1} & \dots & \dots & \dots & a_{mn} \end{pmatrix}$  which is  $(m \times n)$  matrix where

with  $a_{ij}$  elements,  $i = \text{no. of rows}$ ,  $j = \text{no. of Column}$ .

- Real Matrix: with elements are all real.

- Imaginary Matrix: not all elements are real.

- Square Matrix: have the same numbers of Rows & Columns.

-  $(a_1 \ a_2 \ a_3)$  is called row vector of  $(1 \times 3)$  dimension.

-  $\begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}$  is called Column vector of  $(3 \times 1)$  dimension.

- Matrix Addition (subtraction): Matrices should have the same dimension.

Ex:  $A = \begin{pmatrix} 3 & -1 & 2 \\ 5 & 1 & 0 \\ -2 & 1 & -3 \end{pmatrix}$ ,  $B = \begin{pmatrix} 2 & -2 & 1 \\ 3 & 1 & -4 \\ 2 & -1 & 1 \end{pmatrix}$  then  $A+B = \begin{pmatrix} 5 & -3 & 3 \\ 8 & 2 & -4 \\ 0 & 0 & -2 \end{pmatrix}$

- Multiplication of Matrices: Multiplication exists if the inner dimensions are equal, then the resultant matrix is of the outer dimensions.

Ex:  $(3 \times 2)(2 \times 1)$  Ans.  $(3 \times 1)$ ,  $(2 \times 3)(2 \times 3)$  Ans. Multiplication is not valid.

- Transpose of a Matrix: Interchange Rows & Columns.

Ex: if  $A = \begin{pmatrix} 2 & 3 \\ -1 & 2 \\ 3 & 1 \end{pmatrix}$  then  $A^T = \begin{pmatrix} 2 & -1 & 3 \\ 3 & 2 & 1 \end{pmatrix}$

- Symmetric Matrix if  $A = A^T$ , skew matrix if  $A = -A^T$

- Hermitian Matrix if  $A = \bar{A}^T$  skew Hermitian if  $A = -\bar{A}^T$

- Complex Conjugate of a Matrix: if all elements of  $(A)$  are replaced by its Complex conjugate then  $(A')$  is Complex conjugate of  $A$ .

- Unit Matrix (or Identity matrix): if principal diagonal elements are all (1's) and other elements are zeroes. Ex:  $A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$

- Null Matrix: if all elements are zeroes.

- Determinant of a Matrix: for a square matrix there is a determinant

Ex:  $A = \begin{pmatrix} 1 & 2 \\ 4 & 3 \end{pmatrix}$ , then  $|A| = (1 \times 3 - 2 \times 4) = -5$  then  $A$  is called non-singular matrix and invertible (i.e.,  $A^{-1}$  inverse of a matrix is valid).

Ex:  $A = \begin{pmatrix} 1 & -1 \\ 4 & -4 \end{pmatrix}$ , then  $|A| = 0$  and  $A$  is called singular matrix and non-invertible.

- Inverse of a Matrix: if  $A$  is square and non-singular then  $(A^{-1})$  inverse of a matrix is available with  $A^{-1} = \frac{(\text{adjoint coefficient } a_{ij})^T}{\det. \text{ of } A}$

Ex:  $A = \begin{pmatrix} 1 & 2 \\ 4 & 3 \end{pmatrix}$  then  $A^{-1} = \frac{\begin{pmatrix} 3 & -4 \\ -2 & 1 \end{pmatrix}^T}{-5} = \frac{\begin{pmatrix} 3 & -2 \\ -4 & 1 \end{pmatrix}}{-5}$   
then  $A \cdot A^{-1} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$

- Scalar multiplication and division: if a matrix  $A$  is multiplied or divided by scalar value all its elements are multiplied (divided) by this scalar.

Ex:  $A = \begin{pmatrix} 1 & 2 \\ 4 & 3 \end{pmatrix}$  then  $2A = \begin{pmatrix} 2 & 4 \\ 8 & 6 \end{pmatrix}$ .

Ex:  $A = \begin{pmatrix} 3 & -2 \\ -4 & 1 \end{pmatrix}$  then  $A = \begin{pmatrix} \frac{3}{5} & -\frac{2}{5} \\ -\frac{4}{5} & \frac{1}{5} \end{pmatrix}$ .

- Eigen value are the roots of  $(\lambda I - A) = 0$  equation.

\* Why state-space Representation: - "للظلمة"

As a different mathematical Representation compared with that of classical mathematical Representation ( $\frac{O/P}{I/P}$ ) "Transfer Function Model"

There are many reasons to use S.S. Representation. The following is a comparison between the two approaches from the control theories point of views.

S.S. Mathematical Representation	classical (T-F) Math. Representation
1. Is easily applicable for Multi-Input-Multi-output (MIMO) and Single-Input/Single-output (SISO) Systems which may be linear or Non-linear, time invariant (TIV) or (TV).	1. Is easily applicable to linear time-invariant SISO systems.
2. Is essentially a time-domain approach.	2. Is a Complex Frequency-domain approach.
3. Helps to design optimal Control Systems.	3. Some approaches are based on trial and error procedures that will not yield to optimal sys. design.
4. Design will be carried out for a class of I/Ps.	4. Design will be carried out for a specific I/P function.

### \* S.S. Mathematical Representation basic definitions:

1. State: is the smallest variable one can deal with. Ex: 2<sup>nd</sup> order has 2-state.
2. state-space: is the n-dimensional space whose coordinate axes consist of the  $x_1, x_2, \dots, x_n$  axis, and any state can be represented by a point in the state-space.
3. State-space Equations:

$$\dot{\underline{x}} = A\underline{x} + B u \quad (\text{state Equation})$$

$$y = C\underline{x} + D u \quad (\text{output Equation})$$

where;  $\underline{x}$  is the state-vector of (n x 1) elements, where n is the system order.

A is called the dynamic matrix of (n x n) dimension,

B is called the input matrix of (n x 1) dimension.

u is the input for SISA is of (1 x 1) dimension.

y is the output for SISO is of (1 x 1) dimension.

C is the output matrix of (1 x n) dimension.

D is the direct matrix "exists for proper T.F only" with (1 x 1) dimension.

### \* Solution of state Equation $\dot{\underline{x}} = A\underline{x} + B u$

(a) Solution of Homogenous (no-input) state equation  $\dot{\underline{x}} = A\underline{x}$

$$\dot{\underline{x}} = A\underline{x} \Rightarrow \mathcal{L}\{\underline{x}(s) - \underline{x}(0)\} = A\underline{x}(s) \Rightarrow (sI - A)\underline{x}(s) = \underline{x}(0) \quad \underline{x}(t) = e^{At} \underline{x}(0)$$

$$\text{then } \underline{x}(s) = (sI - A)^{-1} \underline{x}(0) \Rightarrow \underline{x}(t) = \mathcal{L}^{-1}\{(sI - A)^{-1} \underline{x}(0)\} \quad \text{then } y(t) = C\underline{x}(t) + D u$$

Note:  $\Phi(t) = e^{At} = \mathcal{L}^{-1}\{(sI - A)^{-1}\}$  = state-transition matrix.

(b) Solution of Non-homogenous state equation  $\dot{\underline{x}} = A\underline{x} + B u$

$$\dot{\underline{x}} = A\underline{x} + B u \Rightarrow \mathcal{L}\{\underline{x}(s) - \underline{x}(0)\} = A\underline{x}(s) + B u(s) \Rightarrow (sI - A)\underline{x}(s) = \underline{x}(0) + B u(s) \Rightarrow$$

$$\underline{x}(s) = (sI - A)^{-1} \underline{x}(0) + (sI - A)^{-1} B u(s) \Rightarrow \underline{x}(t) = \mathcal{L}^{-1}\{(sI - A)^{-1} \underline{x}(0)\} + \mathcal{L}^{-1}\{(sI - A)^{-1} B u(s)\}$$

$$\text{then } \underline{x}(t) = e^{At} \underline{x}(0) + \int_0^t e^{A(t-\tau)} B u(\tau) d\tau, \quad \text{then } y(t) = C\underline{x}(t) + D u. \quad \text{Convolution Integral}$$

\* properties of state-transition matrix  $\Phi(t) = e^{At}$  = exponential matrix =  $\mathcal{L}^{-1}\{(sI - A)^{-1}\}$

$$\Phi(0) = I, \quad \Phi^{-1}(t) = \Phi(-t), \quad \Phi(t_1 + t_2) = \Phi(t_1) \Phi(t_2) = \Phi(t_2) \Phi(t_1), \quad \Phi(t)^n = \Phi(nt),$$

$$\Phi(t_2 - t_1) \Phi(t_1 - t_0) = \Phi(t_2 - t_0) = \Phi(t_1 - t_0) \Phi(t_2 - t_1)$$

$$e^{At} = \sum_{k=0}^{\infty} \frac{A^k t^k}{k!}, \quad \frac{d}{dt} e^{At} = A e^{At} = e^{At} A, \quad e^{At} e^{-At} = I, \quad e^{(A+B)t} = e^{At} e^{Bt} \quad \text{if } AB = BA$$

\* Evaluating  $\frac{O/P}{I/P}$  relation using S.S. Representation:

Since  $\dot{x} = Ax + Bu$  and  $y = Cx + Du$

$$sI \underline{x} - A \underline{x} = Bu \quad \rightarrow \quad y = C(sI - A)^{-1} Bu + Du$$

$$(sI - A) \underline{x} = Bu \quad \rightarrow \quad y = (C(sI - A)^{-1} B + D) u$$

$$\therefore \underline{x} = (sI - A)^{-1} Bu \quad \therefore \frac{O/P}{I/P} = \frac{y(s)}{u(s)} = C(sI - A)^{-1} B + D$$

Note: Signal Flow Graph is one of the ways to extract s-s. equations.

Ex: Given  $y'' + 3y' + 2y = x$  it is required to find its canonical s.s. rep.

check for  $\frac{O/P(s)}{I/P(s)}$  & calculate  $e^{At}$  to find homogenous solution.

solutions:  $s^2 y + 3s y + 2y = x \Rightarrow \frac{y(s)}{x(s)} = \frac{1}{s^2 + 3s + 2}$ , n=2 order

$$\therefore A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix}, B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, C = [1 \ 0], D = 0$$

$$\text{then } \underline{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u, \quad y = [1 \ 0] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + D$$

$$\therefore \dot{x}_1 = x_2, \quad \dot{x}_2 = -2x_1 - 3x_2 + u, \quad y = x_1$$

$$\text{to check: } \frac{y(s)}{u(s)} = C(sI - A)^{-1} B + D = [1 \ 0] \left( \begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix} - \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \right)^{-1} \begin{bmatrix} 0 \\ 1 \end{bmatrix} + 0$$

$$\frac{y(s)}{u(s)} = [1 \ 0] * \begin{pmatrix} s & -1 \\ 2 & s+3 \end{pmatrix}^{-1} * \begin{bmatrix} 0 \\ 1 \end{bmatrix} = [1 \ 0] \begin{pmatrix} s+3 & 1 \\ -2 & s \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \\ s^2 + 3s + 2$$

$$\frac{y(s)}{u(s)} = [1 \ 0] \begin{pmatrix} 1 & \\ & s \end{pmatrix} = \frac{1}{s^2 + 3s + 2}$$

$$\text{to find } e^{At} = \mathcal{L}^{-1} \left\{ (sI - A)^{-1} \right\} = \mathcal{L}^{-1} \left[ \begin{array}{cc} \frac{s+3}{(s+1)(s+2)} & \frac{1}{(s+1)(s+2)} \\ \frac{-2}{(s+1)(s+2)} & \frac{s}{(s+1)(s+2)} \end{array} \right] = \mathcal{L}^{-1} \left[ \begin{array}{cc} \frac{2}{s+1} - \frac{1}{s+2} & \frac{1}{s+1} - \frac{1}{s+2} \\ \frac{-2}{s+1} + \frac{2}{s+2} & \frac{-1}{s+1} + \frac{2}{s+2} \end{array} \right]$$

$$\therefore e^{At} = \begin{bmatrix} 2e^{-t} - e^{-2t} & -e^{-t} + e^{-2t} \\ -2e^{-t} + 2e^{-2t} & -e^{-t} + 2e^{-2t} \end{bmatrix} \quad \text{then } \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} = e^{At} \begin{bmatrix} x_1(0) \\ x_2(0) \end{bmatrix} \text{ homogenous state Eq. solution}$$

then  $y(t) = x_1(t)$  system o/p solution.

Question: if non-homogenous solution is required for  $u = \text{unit step} = 1$ , what will the solution be?

\* Methods for finding  $\Phi(t) = e^{At} = \mathcal{L}^{-1} (sI - A)^{-1}$

1. Diagonalization: if  $A$  is diagonal [ex:  $\begin{bmatrix} 1 & 0 \\ 0 & -2 \end{bmatrix}$ ] then  $e^{At} = \begin{bmatrix} e^{t} & 0 \\ 0 & e^{-2t} \end{bmatrix}$  which is easy even for higher order, but transformation of  $A, B, C, D$  into diagonal wise will not be an easy job. [ $A$  "diagonal" =  $P^{-1}AP$ ,  $B = P^{-1}B$ ,  $C = C'P$ ] i.e. one should find  $P$ -matrix to diagonalize. "غير مضمونة"

2. If the D.E of the system is available, Canonical S.S. Representation form is easy to extract, then one should find  $e^{At}$ . "مضمونة"

3. If one analyse a certain system using state-Representation a Random form of S.S. Representation will be obtained, then  $e^{At}$  should be calculated. "مضمونة"  
Any way for 3-point above  $A, B, C, D$  matrices will be set, and  $e^{At}$  should be calculated either by  $\mathcal{L}^{-1}(sI - A)^{-1}$ , Cayley-Hamilton theorem, Sylvester's criterion, and Fadeev algorithm.

\* Cayley-Hamilton Theorem for finding  $\Phi(t) = e^{At}$  :-

Here, mainly to compute a polynomial  $f(A) = \alpha_0 I + \alpha_1 A + \alpha_2 A^2 + \dots$

where,  $\alpha_0, \alpha_1, \alpha_2, \dots$  are found from function  $g_n(\lambda) = \alpha_0 + \alpha_1 \lambda + \alpha_2 \lambda^2 + \dots$

with  $\lambda$ 's are the eigen values of matrix  $A$ .

Ex:- for  $A = \begin{pmatrix} 0 & 1 \\ 1 & 2 \end{pmatrix}$ , find  $A^{100}$  using Cayley-Hamilton theorem.

$$\begin{array}{ccc} & f(A) & \\ & \swarrow \quad \searrow & \\ f(\lambda_1) = g(\lambda_1) & & g(\lambda_1) = \alpha_0 + \alpha_1 \lambda_1 \quad \text{"less by 1. from } A \text{ order"} \\ f(\lambda_2) = g(\lambda_2) & & g(\lambda_2) = \alpha_0 + \alpha_1 \lambda_2 \end{array}$$

Solve for  $\alpha_0, \alpha_1$ , then  $f(A) = \alpha_0 I + \alpha_1 A$ .

$$(\lambda I - A) = 0 \Rightarrow \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} - \begin{bmatrix} 0 & 1 \\ 1 & 2 \end{bmatrix} = 0 \Rightarrow \begin{bmatrix} \lambda & -1 \\ -1 & \lambda - 2 \end{bmatrix} = 0 \Rightarrow \lambda^2 - 2\lambda - 1 = 0$$

$$\lambda_1 = 1 + \sqrt{2}, \lambda_2 = 1 - \sqrt{2}, f(\lambda_1) = g(\lambda_1) = \alpha_0 + \alpha_1 \lambda_1 \Rightarrow (1 + \sqrt{2})^{100} = \alpha_0 + \alpha_1 (1 + \sqrt{2}) \dots (1)$$

$$f(\lambda_2) = g(\lambda_2) = \alpha_0 + \alpha_1 \lambda_2 \Rightarrow (1 - \sqrt{2})^{100} = \alpha_0 + \alpha_1 (1 - \sqrt{2}) \dots (2)$$

Solve (1) & (2) for  $\alpha_0, \alpha_1$ , then  $A^{100} = f(A) = \alpha_0 I + \alpha_1 A$ .