<u>Lecture – 8</u> State Space Representation of Dynamical Systems

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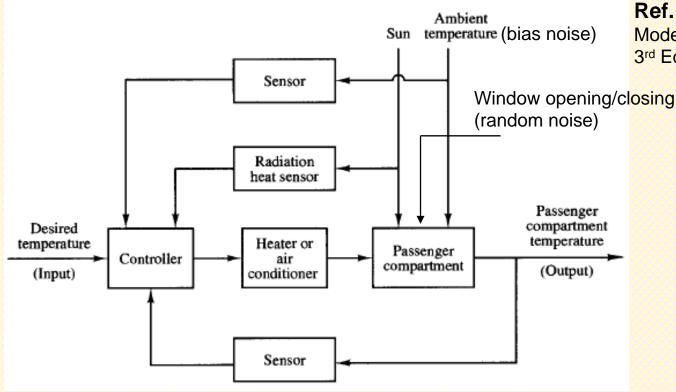
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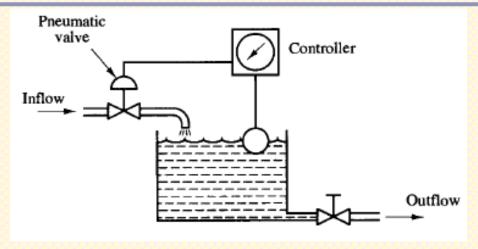
A Practical Control System



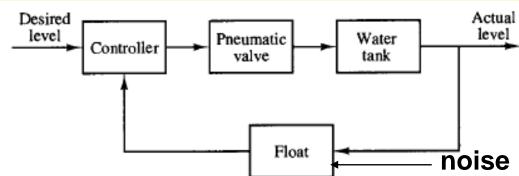
Ref.: K. Ogata, Modern Control Engineering 3rd Ed., Prentice Hall, 1999.

Temperature control system in a car

Another Practical Control System



Ref.: K. Ogata, Modern Control Engineering 3rd Ed., Prentice Hall, 1999.



Water level control in an overhead tank

State Space Representation

Input variable:

- Manipulative (control)
- Non-manipulative (noise)

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Output variable:

Variables of interest that can be either be measured or calculated

State variable:

Minimum set of parameters which completely summarize the system's status.

Definitions

State: The state of a dynamic system is the smallest number of variables (called **state variables**) such that the knowledge of these variables at $t = t_0$, together with the knowledge of the input for $t = t_0$, completely determine the behaviour of the system for any time $t \ge t_0$.

Note: State variables need not be physically measurable or observable quantities. This gives extra flexibility.

Definitions

State Vector: A n - dimensional vector whos components are n state variables that describe the system completely.

State Space: The n-dimensional space whose co-ordinate axes consist of the x_1 axis, x_2 axis, \dots , x_n axis is called a state space.

Note: For any dynamical system, the state space remains unique, but the state variables are not unique.

Critical Considerations while Selecting State Variables

- Minimum number of variables
 - Minimum number of first-order differential equations needed to describe the system dynamics completely
 - Lesser number of variables: won't be possible to describe the system dynamics
 - Larger number of variables:
 - Computational complexity
 - Loss of either controllability, or observability or both.
- Linear independence. If not, it may result in:
 - Bad: May not be possible to solve for all other system variables
 - Worst: May not be possible to write the complete state equations

State Variable Selection

- Typically, the number of state variables (*i.e.* the order of the system) is equal to the number of independent energy storage elements. However, there are exceptions!
- Is there a restriction on the selection of the state variables?
 - **YES!** All state variables should be linearly independent and they must collectively describe the system completely.

Advantages of State Space Representation

- Systematic analysis and synthesis of higher order systems without truncation of system dynamics
- Convenient tool for MIMO systems
- Uniform platform for representing time-invariant systems, time-varying systems, linear systems as well as nonlinear systems
- Can describe the dynamics in almost all systems (mechanical systems, electrical systems, biological systems, economic systems, social systems etc.)
- Note: Transfer function representations are valid for only for linear time invariant (LTI) systems

Generic State Space Representation

$$X \triangleq \begin{bmatrix} x_1 & \cdots & x_n \end{bmatrix}^T \in \mathbb{R}^n, \quad U \triangleq \begin{bmatrix} u_1 & \cdots & u_m \end{bmatrix}^T \in \mathbb{R}^m$$

$$\begin{bmatrix} \dot{x}_1 \\ \vdots \\ \dot{x}_n \end{bmatrix} = \begin{bmatrix} f_1(t, x_1 & \cdots & x_n, u_1 & \cdots & u_m) \\ \vdots & \vdots & & & \\ f_n(t, x_1 & \cdots & x_n, u_1 & \cdots & u_m) \end{bmatrix}, \quad t \in \mathbb{R}^+$$

Generic State Space Representation

$$Y \triangleq \begin{bmatrix} y_1 & \cdots & y_p \end{bmatrix}^T \in \mathbb{R}^p$$

$$\begin{bmatrix} y_1 \\ \vdots \\ y_p \end{bmatrix} = \begin{bmatrix} h_1(t, x_1 & \cdots & x_n, u_1 & \cdots & u_m) \\ \vdots & \vdots & & & \\ h_p(t, x_1 & \cdots & x_n, u_1 & \cdots & u_m) \end{bmatrix}, \quad t \in \mathbb{R}^+$$

$$h(t, X, U)$$

Summary:
$$\dot{X} = f(t, X, U)$$
: A set of differential equations

Y = h(t, X, U): A set of algebraic equations

State Space Representation (noise free systems)

Nonlinear System

$$\dot{X} = f(X, U)$$

$$Y = h(X, U)$$

$X \in \mathbb{R}^n, U \in \mathbb{R}^m$

$$Y \in \mathbb{R}^p$$

Linear System

$$\dot{X} = AX + BU$$

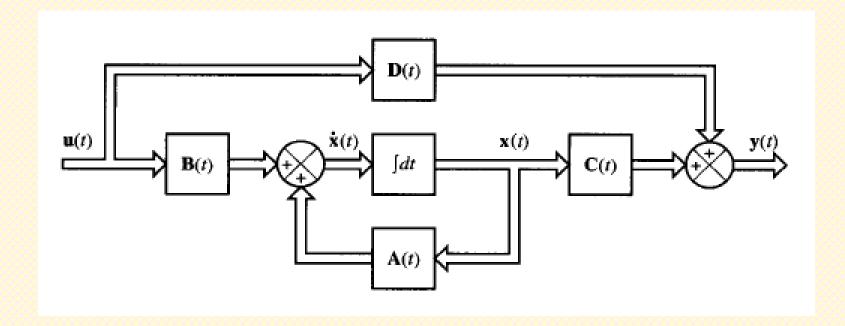
$$Y = CX + DU$$

$$B$$
 - Input matrix- $n \times m$

$$C$$
 - Output matrix- $p \times n$

$$D$$
 - Feed forward matrix – $p \times m$

Block diagram representation of linear systems



Ref.: K. Ogata, Modern Control Engineering 3rd Ed., Prentice Hall, 1999.

Writing Differential Equations in First Companion Form

(Phase variable form/Controllable canonical form)

$$\frac{d^{n}y}{dt^{n}} + a_{n-1}\frac{d^{n-1}y}{dt^{n-1}} + \cdots + a_{0}y = u$$

Choose output y(t) and its (n-1) derivatives as state variables

$$\begin{bmatrix} x_1 = y \\ x_2 = \frac{dy}{dt} \\ \vdots \\ x_n = \frac{d^{n-1}y}{dt^{n-1}} \end{bmatrix}$$

$$\dot{x}_{1} = \frac{dy}{dt}$$

$$\dot{x}_{2} = \frac{d^{2}y}{dt^{2}}$$

$$\vdots$$

$$\dot{x}_{n} = \frac{d^{n}y}{dt^{n}}$$

First Companion Form

(Controllable Canonical Form)

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \\ \vdots \\ \dot{x}_{n-1} \\ \dot{x}_n \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & \cdots & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & \cdots & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & \cdots & 0 & 0 \\ \vdots & \vdots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \\ -a_0 & -a_1 & -a_2 & \cdots & \cdots & \cdots & -a_{n-1} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_{n-1} \\ x_n \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} u$$

$$y = \begin{bmatrix} 1 & 0 & 0 & \cdots & \cdots & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_{n-1} \\ x_n \end{bmatrix} + \begin{bmatrix} 0 \end{bmatrix} u$$

Example - 1

• Dynamical system $\ddot{x} + 3\dot{x} + 2x = u$ y = x

• State variables:
$$x_1 \triangleq x$$
, $x_2 \triangleq \dot{x}$

$$\dot{x}_1 = x_2, \quad \dot{x}_2 = -2x_1 - 3x_2 + u$$
Hence $\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$

$$y = \underbrace{\begin{bmatrix} 1 & 0 \end{bmatrix}}_{C} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \underbrace{\begin{bmatrix} 0 \end{bmatrix}}_{D} u$$

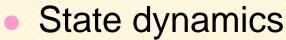
Example – 2 (spring-mass-damper system)

System dynamics

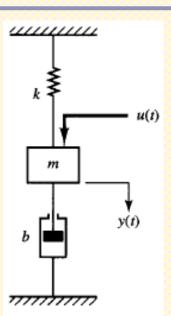
$$m\ddot{y} + c\dot{y} + ky = bu$$

State variables

$$x_1 \triangleq y, \quad x_2 \triangleq \dot{y}$$



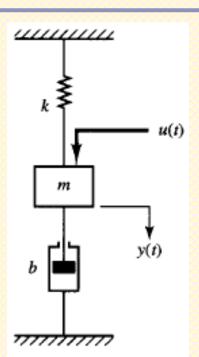
$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} \dot{y} \\ \frac{1}{m}(-ky - c\dot{y}) + \frac{b}{m}u \end{bmatrix} = \begin{bmatrix} x_2 \\ \frac{1}{m}(-kx_1 - cx_2) + \frac{b}{m}u \end{bmatrix}$$



Example - 2 (spring-mass-damper system)

State dynamics

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -k & -c \\ m & m \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ b \\ m \end{bmatrix} u$$

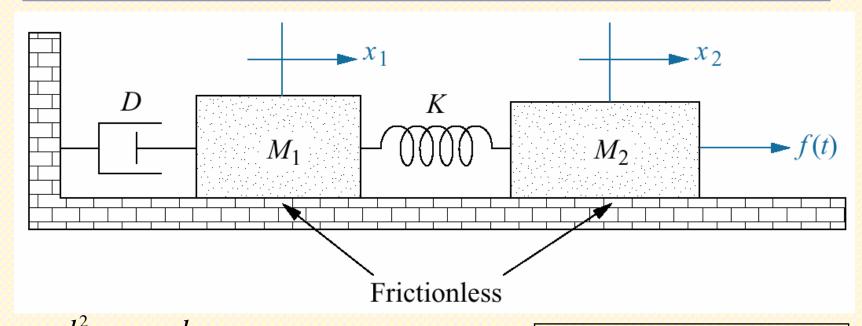


Output equation

$$y \triangleq x_1$$

$$y = \begin{bmatrix} 1 & \cdots & 0 \end{bmatrix} X + \begin{bmatrix} 0 \end{bmatrix} u$$

Example - 3: Translational Mechanical System



$$M_{1} \frac{d^{2}x_{1}}{dt^{2}} + D \frac{dx_{1}}{dt} + K(x_{1} - x_{2}) = 0$$

$$M_{2} \frac{d^{2}x_{2}}{dt^{2}} + K(x_{2} - x_{1}) = f(t)$$

Ref: N. S. Nise: Control Systems Engineering, 4th Ed., Wiley, 2004

Example - 3: Translational Mechanical System

Define

$$v_1 \triangleq \frac{dx_1}{dt}, \quad v_2 \triangleq \frac{dx_2}{dt}$$

System equations

$$\frac{dv_1}{dt} = -\frac{K}{M_1} x_1 - \frac{D}{M_1} v_1 + \frac{K}{M_1} x_2$$

$$\frac{dv_2}{dt} = \frac{K}{M_2} x_1 - \frac{K}{M_2} x_2 + \frac{1}{M_2} f(t)$$

State space equations in standard form

$$\begin{bmatrix} \dot{x}_1 \\ \dot{v}_1 \\ \dot{x}_2 \\ \dot{v}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ -\frac{K}{M_1} & -\frac{D}{M_1} & \frac{K}{M_1} & 0 \\ 0 & 0 & 0 & 1 \\ \frac{K}{M_2} & 0 & -\frac{K}{M_2} & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ v_1 \\ x_2 \\ v_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \\ \frac{1}{M_2} \end{bmatrix} f(t)$$

Example - 4: Nonlinear spring in Example - 3

• Dynamic equations $M_1 \frac{d^2 x_1}{dt^2} + D \frac{dx_1}{dt} + K(x_1 - x_2)^3 = 0$ $M_2 \frac{d^2 x_2}{dt^2} + K(x_2 - x_1)^3 = f(t)$

State space equation

$$\dot{x}_1 = v_1$$

$$\dot{v}_1 = -\frac{K}{M_1} (x_1 - x_2)^3 - \frac{D}{M_1} v_1$$

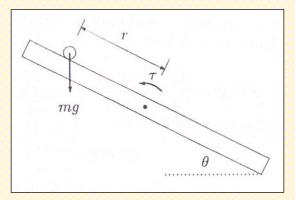
$$\dot{x}_2 = v_2$$

$$\dot{v}_2 = -\frac{K}{M_2} (x_2 - x_1)^3 + \frac{1}{M_2} f(t)$$

Example - 5 The Ball and Beam System

The beam can rotate by applying a torque at the centre of rotation, and ball can move freely along

the beam



Moment of Inertia of beam: J

Mass, moment of inertia and radius of ball: m, J_b, R

Example - 5 The Ball and Beam System

State Space Model

$$x_1 = r, x_2 = \dot{r}, x_3 = \theta, x_4 = \dot{\theta}$$

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = \frac{-mg\sin x_3 + mx_1x_4^2}{m + \frac{J_b}{R^2}}$$

$$\dot{x}_3 = x_4$$

$$\dot{x}_4 = \frac{\tau - mg \ x_1 \cos x_3 - 2m \ x_1 x_2 x_4}{m \ x_1^2 + J + J_b}$$

Example - 6: Van-der Pol's Oscillator (Limit cycle behaviour)

- Equation $M \ddot{x} + 2c(x^2 1)\dot{x} + k x = 0$ $\{c, k > 0\}$

- State variables $x_1 \triangleq x$, $x_2 \triangleq \dot{x}$

State Space Equation

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \underbrace{\begin{bmatrix} x_2 \\ -\frac{2c}{m} (x_1^2 - 1) x_2 - \frac{k}{m} x_1 \end{bmatrix}}_{F(X)}$$
: Homogeneous nonlinear system

Example - 7: Spinning Body Dynamics (Satellite dynamics)

Dynamics:

$$\dot{\omega}_1 = \left(\frac{I_2 - I_3}{I_1}\right) \omega_2 \omega_3 + \left(\frac{1}{I_1}\right) \tau_1$$

$$\dot{\omega}_2 = \left(\frac{I_3 - I_1}{I_2}\right) \omega_3 \omega_1 + \left(\frac{1}{I_2}\right) \tau_2$$

$$\dot{\omega}_3 = \left(\frac{I_1 - I_2}{I_3}\right) \omega_1 \omega_2 + \left(\frac{1}{I_3}\right) \tau_3$$

 I_1, I_2, I_3 : MI about principal axes

 $\omega_1, \omega_2, \omega_3$: Angular velocities about principal axes

 τ_1, τ_2, τ_3 : Torques about principal axes

Example - 8: Airplane Dynamics, Six Degree-of-Freedom Nonlinear Model

Ref: Roskam J., Airplane Flight Dynamics and Automatic Controls, 1995

$$\dot{U} = VR - WQ - g\sin\Theta + \left(F_{A_X} + F_{T_X}\right)/m$$

$$\dot{V} = WP - UR + g \sin \Phi \cos \Theta + (F_{A_Y} + F_{T_Y})/m$$

$$\dot{W} = UQ - VP + g\cos\Phi\cos\Theta + (F_{A_z} + F_{T_z})/m$$

$$\dot{P} = c_1 QR + c_2 PQ + c_3 (L_A + L_T) + c_4 (N_A + N_T)$$

$$\dot{Q} = c_5 PR - c_6 (P^2 - R^2) + c_7 (M_A + M_T)$$

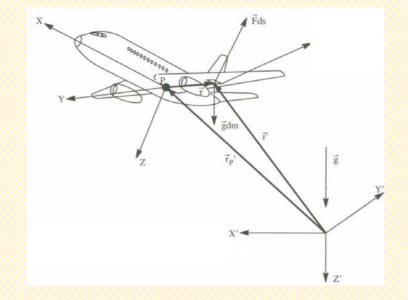
$$\dot{R} = c_8 PQ - c_2 QR + c_4 (L_A + L_T) + c_9 (N_A + N_T)$$

$$\dot{\Phi} = P + Q\sin\Phi\tan\Theta + R\cos\Phi\tan\Theta$$

$$\dot{\Theta} = Q\cos\Phi - R\sin\Phi$$

$$\dot{\Psi} = (Q\sin\Phi + R\cos\Phi)\sec\Theta$$

$$\begin{bmatrix} \dot{X}' \\ \dot{Y}' \\ \dot{Z}' \end{bmatrix} = \begin{bmatrix} \cos \Psi & -\sin \Psi & 0 \\ \sin \Psi & \cos \Psi & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos \Theta & 0 & \sin \Theta \\ 0 & 1 & 0 \\ -\sin \Theta & 0 & \cos \Theta \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \Phi & -\sin \Phi \\ 0 & \sin \Phi & \cos \Phi \end{bmatrix} \begin{bmatrix} U \\ V \\ W \end{bmatrix}$$
[Note: $\dot{h} = -\dot{Z}'$]



Example - 8: Airplane Dynamics, Six Degree-of-Freedom Nonlinear Model

Ref: Roskam J., Airplane Flight Dynamics and Automatic Controls, 1995

$$F_{T_x} = \sum_{i=1}^{N} T_i \cos \Phi_{T_i} \cos \Psi_{T_i} \qquad L_T = -\sum_{i=1}^{N} \left(T_i \cos \Phi_{T_i} \sin \Psi_{T_i} \right) z_{T_i} - \sum_{i=1}^{N} \left(T_i \sin \Phi_{T_i} \right) y_{T_i}$$

$$F_{T_y} = \sum_{i=1}^{N} T_i \cos \Phi_{T_i} \sin \Psi_{T_i} \qquad M_T = \sum_{i=1}^{N} \left(T_i \cos \Phi_{T_i} \cos \Psi_{T_i} \right) z_{T_i} + \sum_{i=1}^{N} \left(T_i \sin \Phi_{T_i} \right) x_{T_i} \qquad T(\alpha) \triangleq \begin{bmatrix} \cos \alpha & -\sin \alpha \\ \sin \alpha & \cos \alpha \end{bmatrix}$$

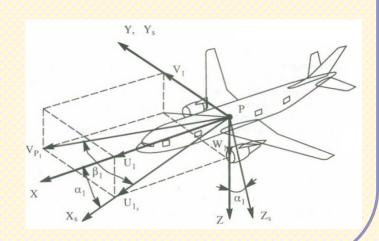
$$F_{T_z} = -\sum_{i=1}^{N} T_i \sin \Phi_{T_i} \qquad N_T = -\sum_{i=1}^{N} \left(T_i \cos \Phi_{T_i} \cos \Psi_{T_i} \right) y_{T_i} + \sum_{i=1}^{N} \left(T_i \cos \Phi_{T_i} \sin \Psi_{T_i} \right) x_{T_i}$$

$$\begin{bmatrix} F_{A_{X}} \\ F_{A_{Z}} \end{bmatrix} = T(\alpha) \begin{bmatrix} F_{A_{X_{s}}} \\ F_{A_{Z_{s}}} \end{bmatrix} = T(\alpha) (-\overline{q}S) \begin{bmatrix} C_{D_{0}} & C_{D_{\alpha}} & C_{D_{i_{h}}} \\ C_{L_{0}} & C_{L_{\alpha}} & C_{L_{i_{h}}} \end{bmatrix} \begin{bmatrix} 1 \\ \alpha \\ i_{h} \end{bmatrix} + \begin{bmatrix} C_{D_{\delta_{E}}} \\ C_{L_{\delta_{E}}} \end{bmatrix} \delta_{E}$$

$$\begin{bmatrix} L_{A} \\ N_{A} \end{bmatrix} = T(\alpha) \begin{bmatrix} L_{A_{S}} \\ N_{A_{S}} \end{bmatrix} = T(\alpha) \overline{q}Sb \begin{pmatrix} C_{l_{\beta}} \\ C_{n_{\beta}} \end{pmatrix} \beta + \begin{bmatrix} C_{l_{\delta_{A}}} & C_{l_{\delta_{R}}} \\ C_{n_{\delta_{A}}} & C_{n_{\delta_{R}}} \end{bmatrix} \begin{bmatrix} \delta_{A} \\ \delta_{R} \end{bmatrix}$$

$$F_{A_{Y}} = \overline{q}S \ C_{Y} = \overline{q}S \left(C_{Y_{\beta}}\beta + \begin{bmatrix} C_{Y_{\delta_{A}}} & C_{Y_{\delta_{R}}} \end{bmatrix} \begin{bmatrix} \delta_{A} \\ \delta_{R} \end{bmatrix} \right)$$

$$M_A = \overline{q}S\overline{c} \ C_m = \overline{q}S\overline{c} \begin{bmatrix} C_{m_o} & C_{m_{i_h}} \end{bmatrix} \begin{bmatrix} 1 & \alpha & i_h \end{bmatrix}^T + C_{m_{\delta_E}} \delta_E$$



Advantages of State Space Representation

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